

# Monitoring Toolbox V1.2 Quick Start Guide

Note: The current version will expire on December 31, 2021. New versions will be released.



### **Objective of this toolbox**

 Monitor the operation of industrial processes, and detect the occurrence of events (e.g., faults, near faults, shutdown) based on process data

### **Key features of this toolbox**

- Five causal analysis algorithms are provided to investigate the cause-effect relationships among process variables
- Five unsupervised learning algorithms are provided to conduct predictive monitoring of the process and to detect events taking place in processes
- Online monitoring feature is provided to simulate the online implementation of monitoring schemes for industrial processes
- Hierarchical distributed monitoring is provided to account for larger processes of more process variables

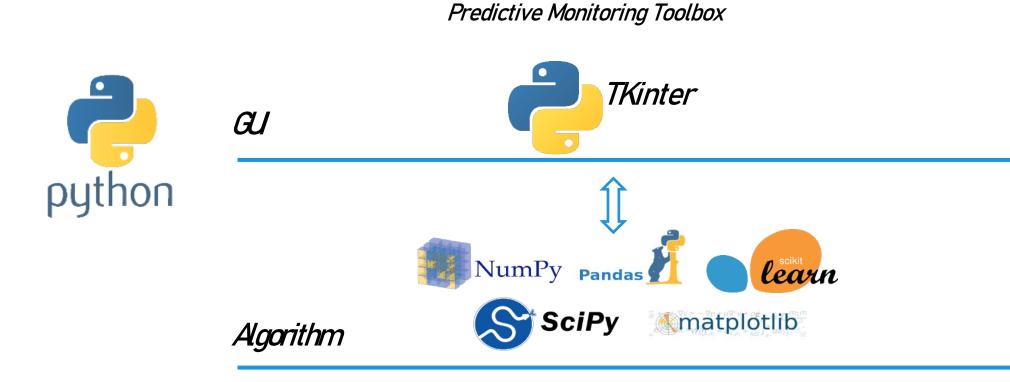


# Overview of Predictive Monitoring Toolbox



#### An overview of the toolbox

The toolbox was developed using Python

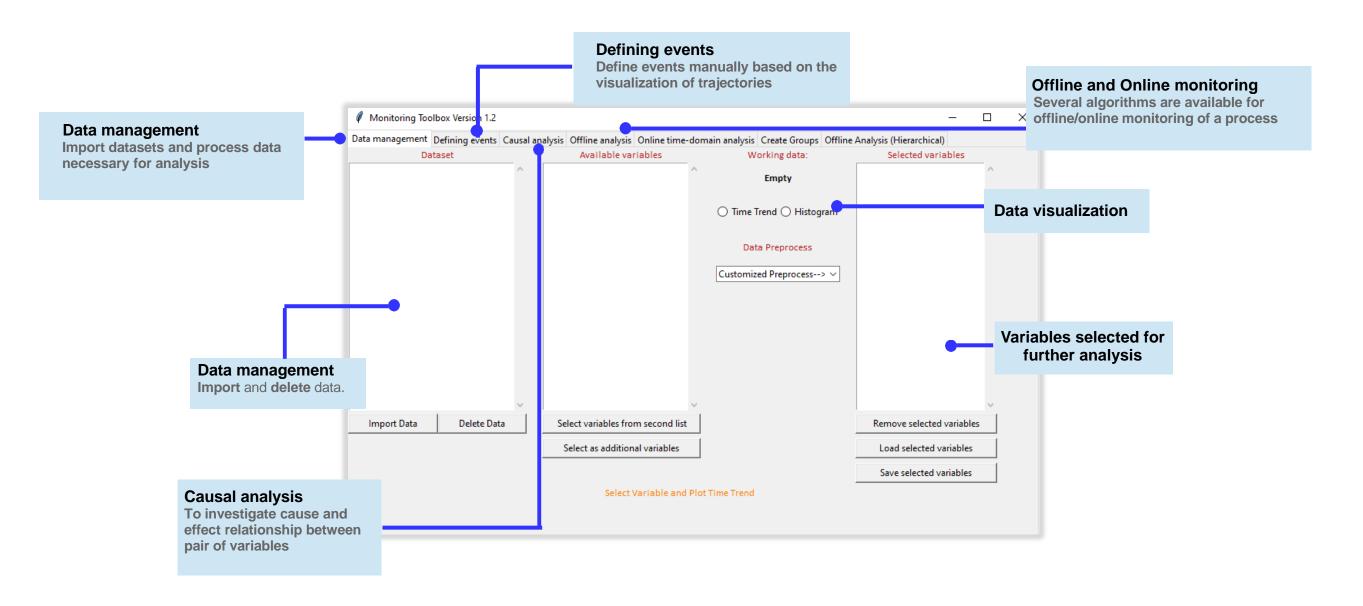


- Python has open source license (free to use)
- Python possesses stable libraries
- Python is developer friendly software



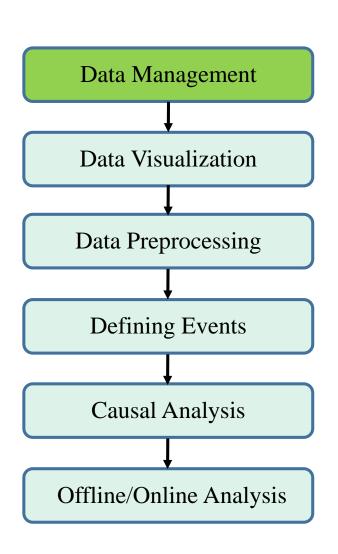
#### An overview of the toolbox

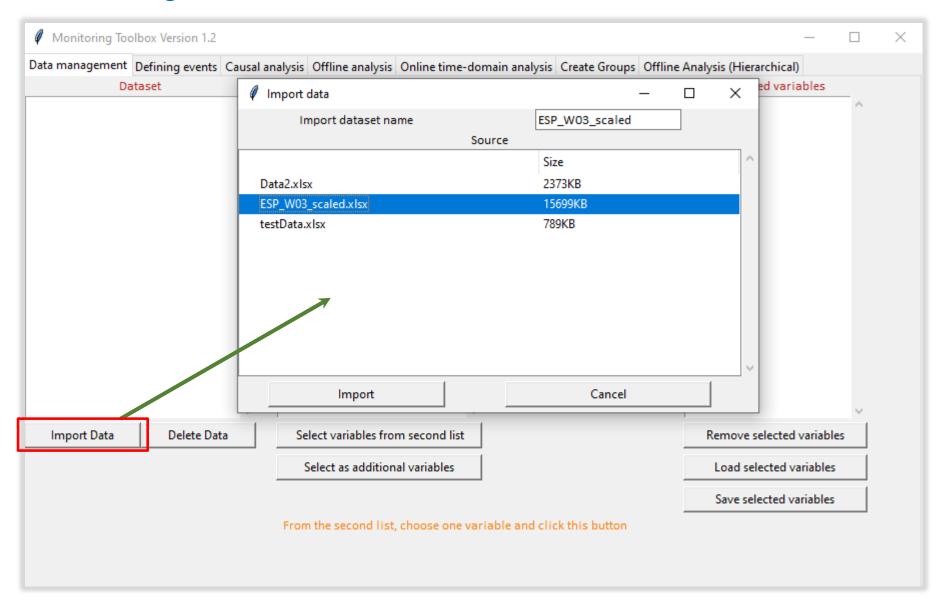
We create this guide to help you minimize the learning curve.





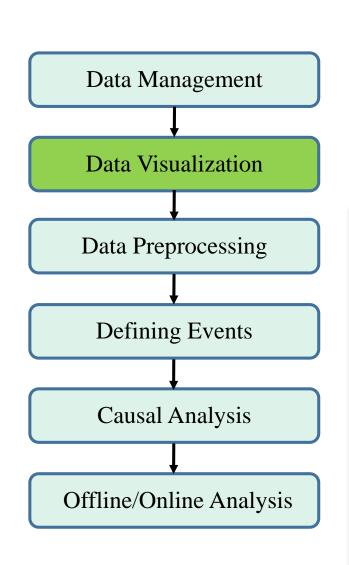
### An overview of the toolbox – Data management

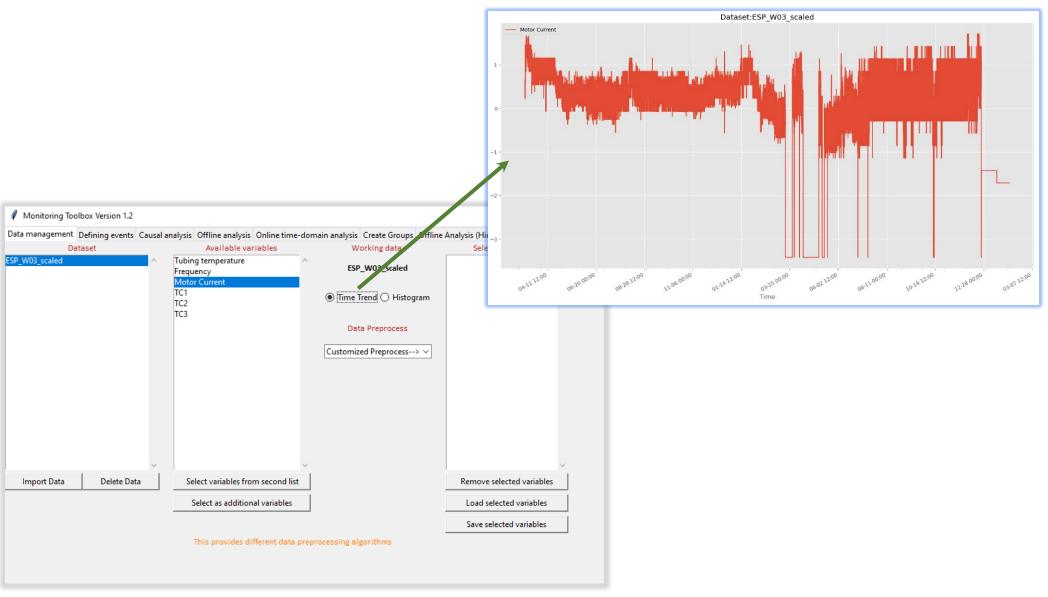






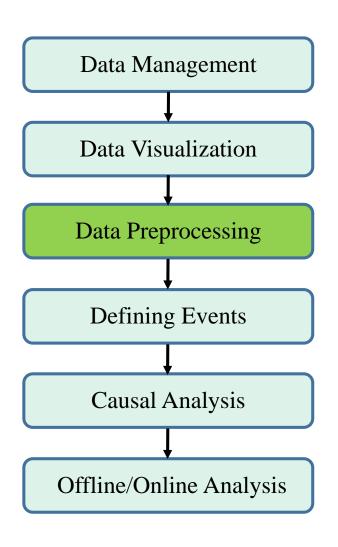
#### An overview of the toolbox – Data visualization

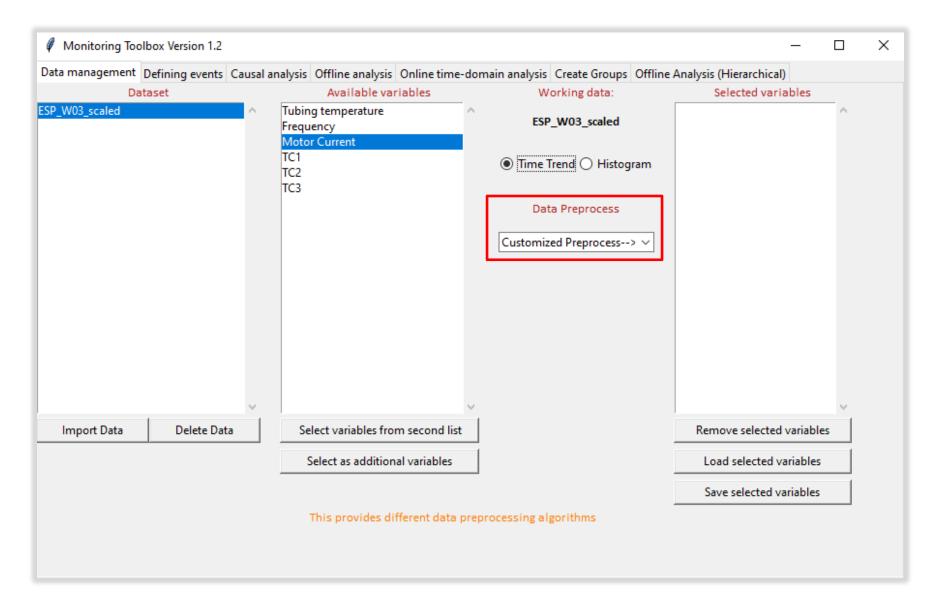






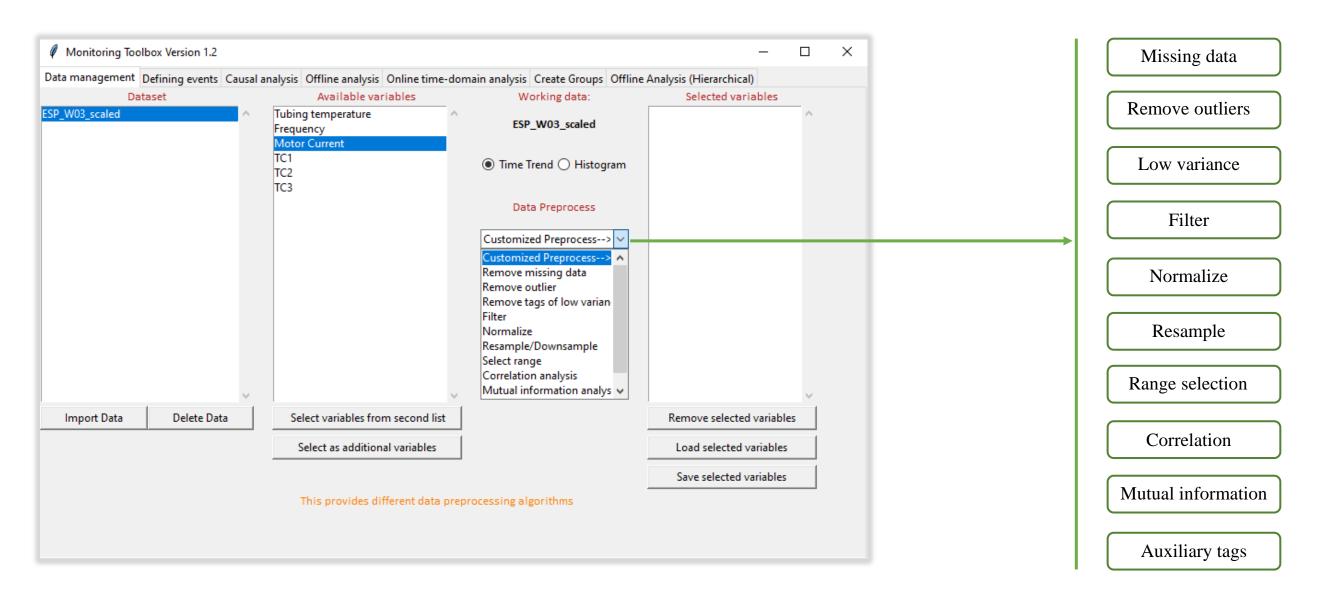
### An overview of the toolbox – Data preprocessing





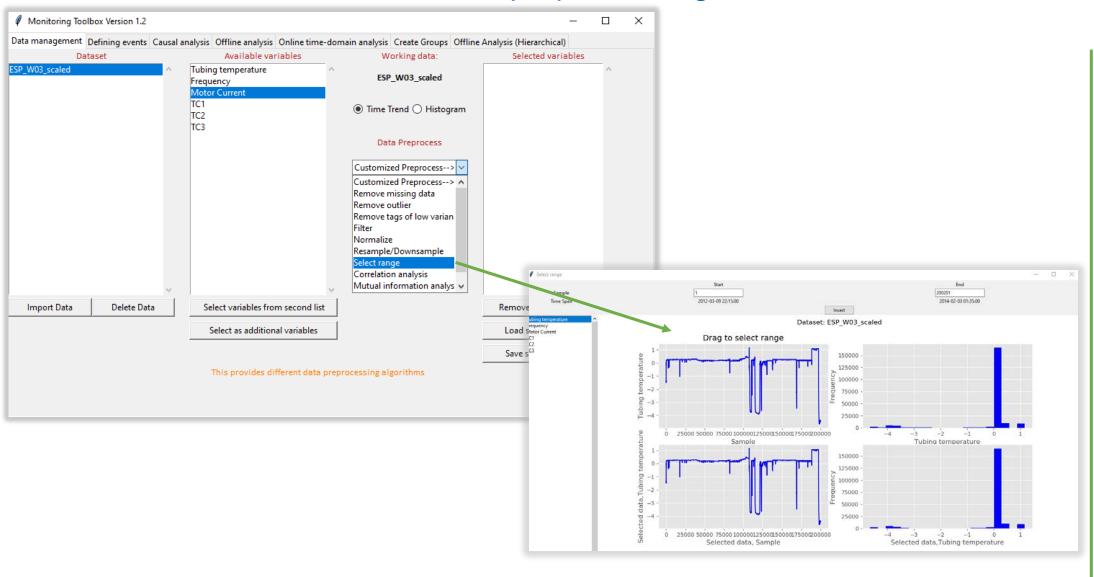


### An overview of the toolbox – Data preprocessing





### An overview of the toolbox – Data preprocessing



Missing data

Remove outliers

Low variance

Filter

Normalize

Resample

Range selection

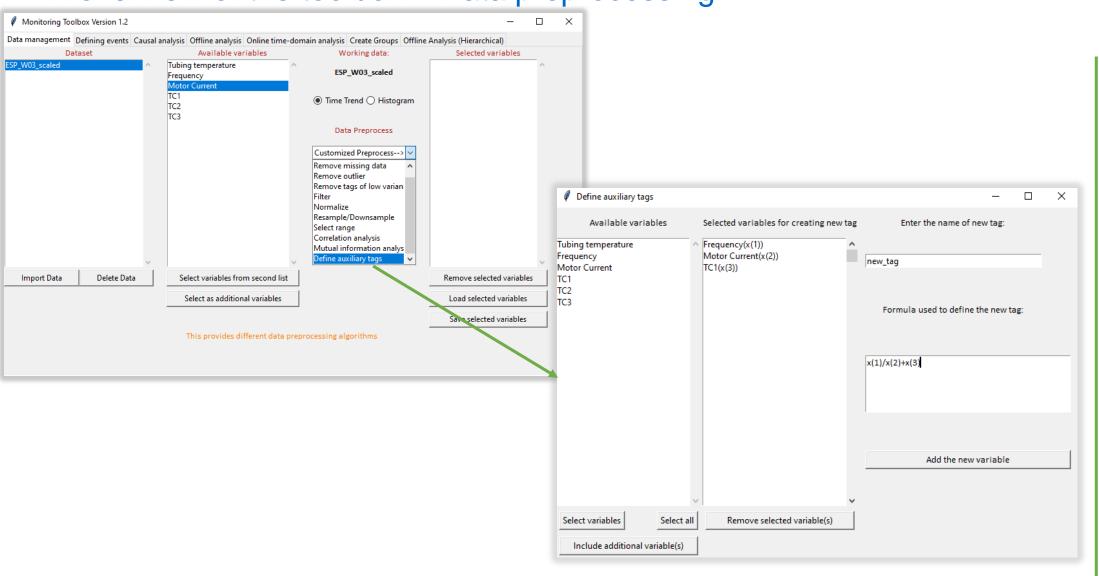
Correlation

Mutual information

Auxiliary tags



### An overview of the toolbox – Data preprocessing



Missing data

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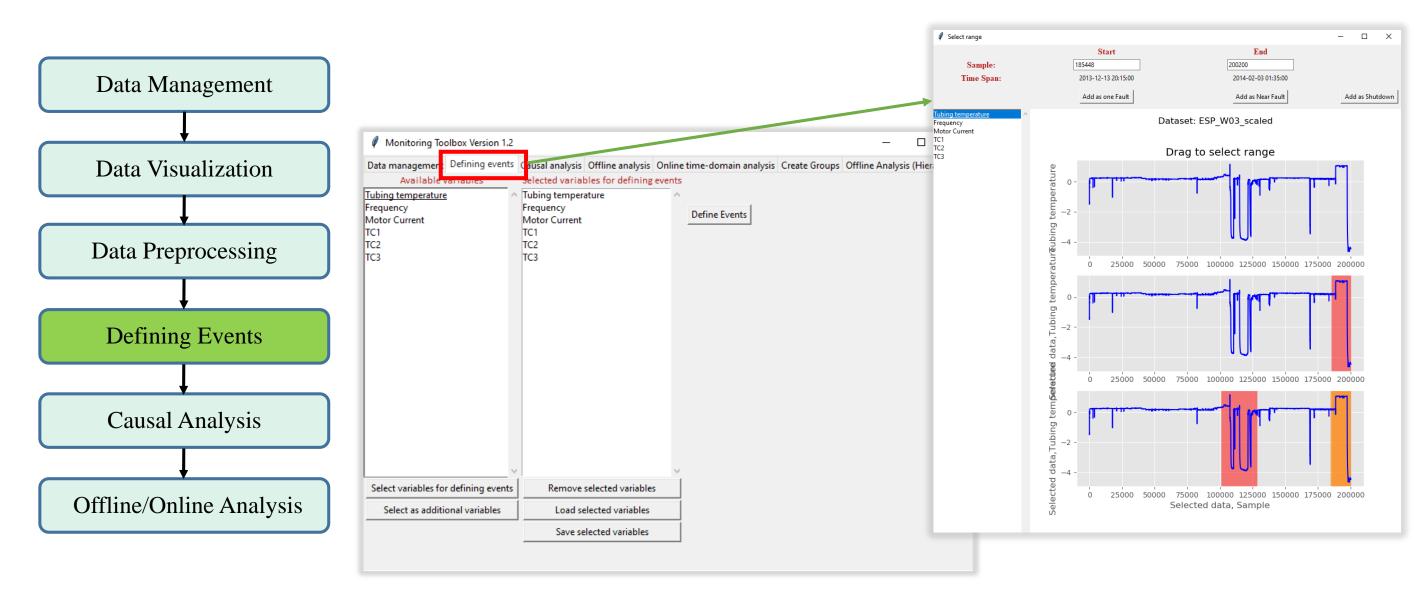
Correlation

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Auxiliary tags

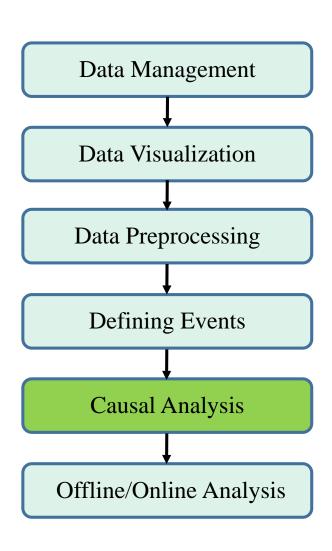


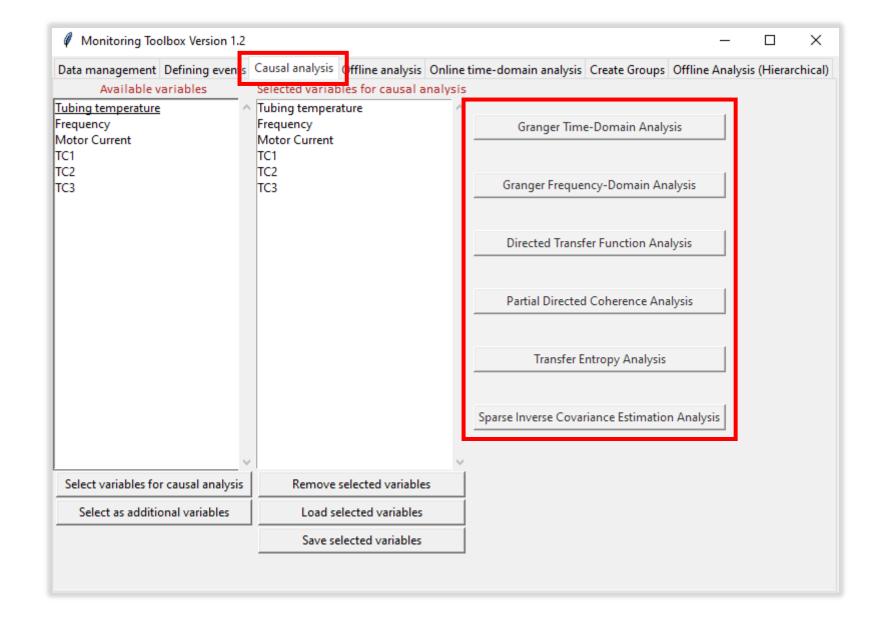
### An overview of the toolbox – Defining events





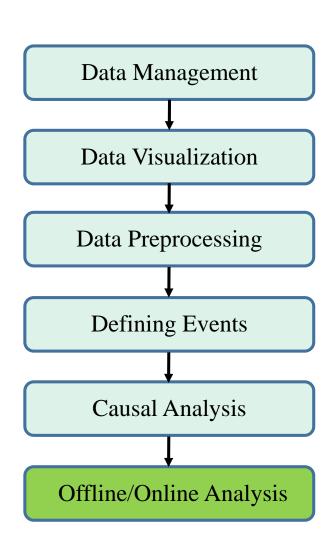
### An overview of the toolbox – Causal analysis

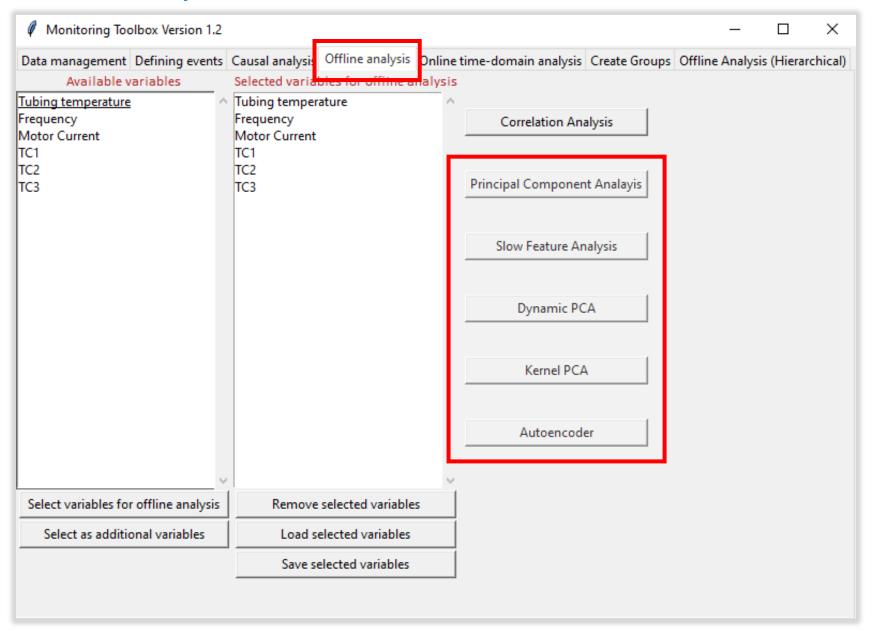






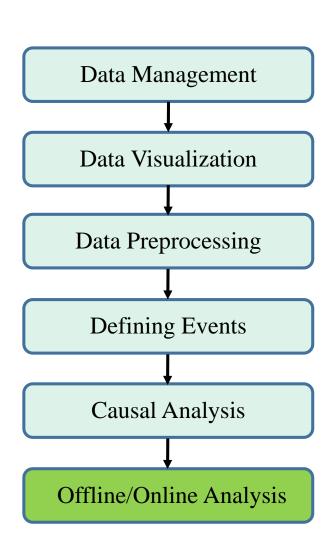
### An overview of the toolbox – Offline analysis

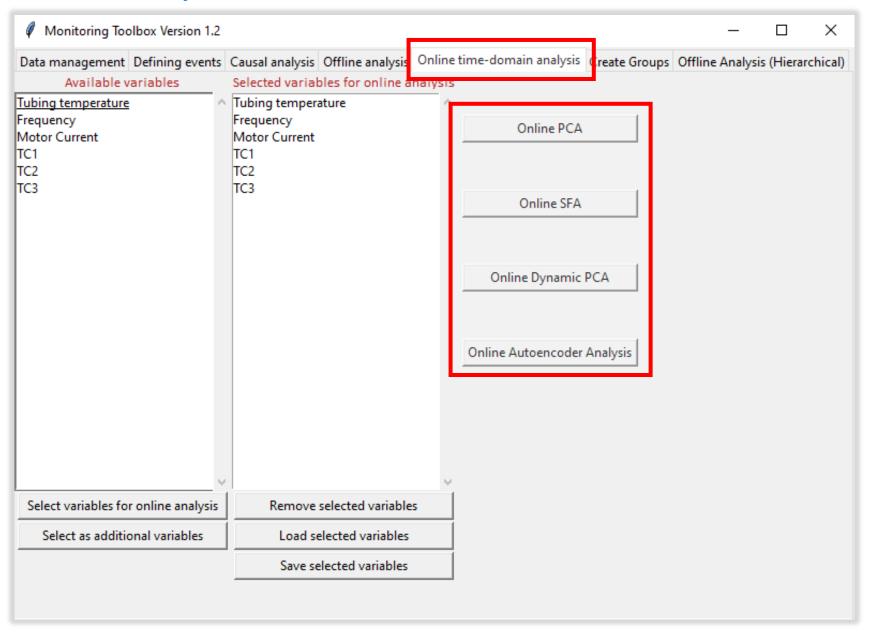






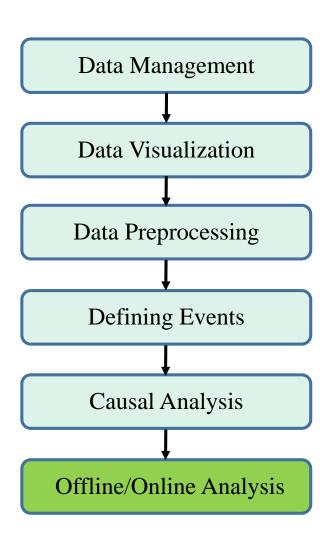
### An overview of the toolbox – Online analysis

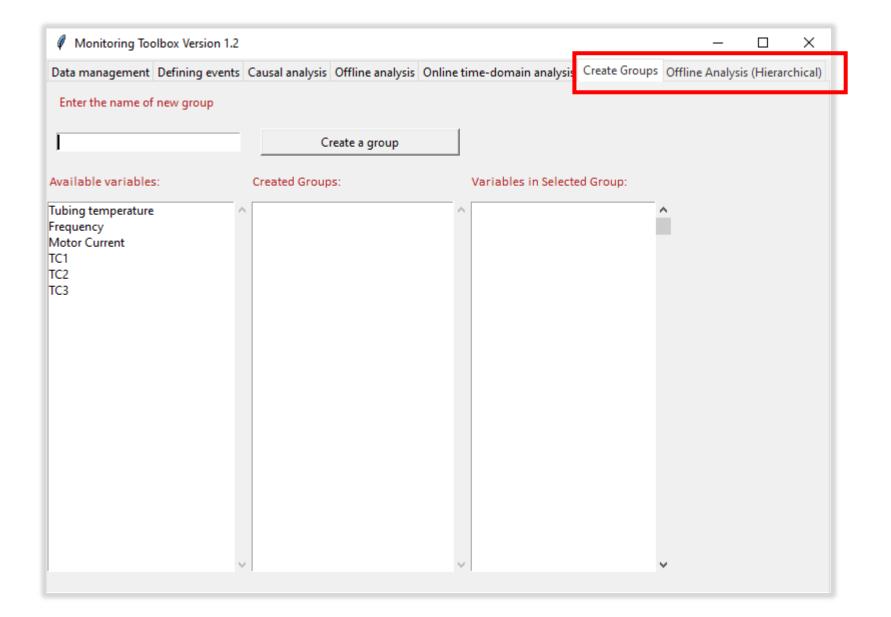






### An overview of the toolbox – Hierarchical distributed analysis





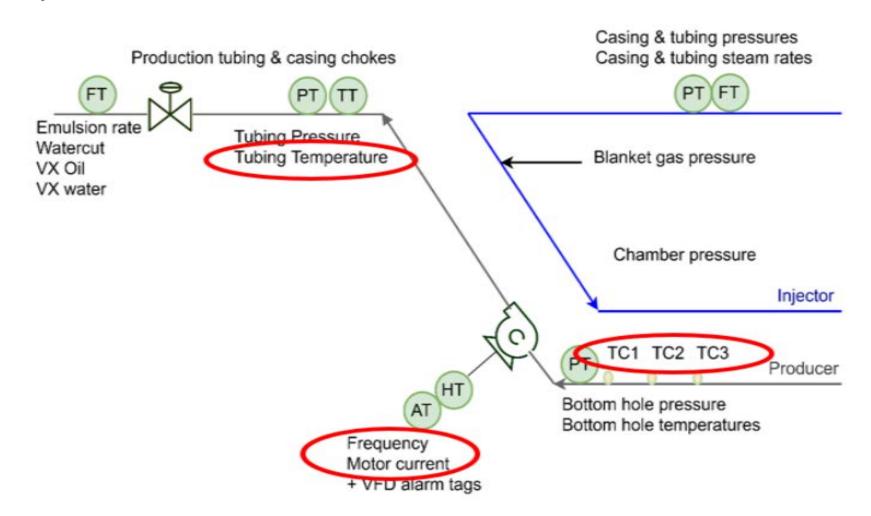


# Start Guide based on a Case Study



### Quick Start Guide – A Case Study on Electrical Submersible Pump (ESP)

- Abnormal events took place in this process.
- The objective is to detect the occurrence of these events based on the process data.
- This case study is based on normalized data for ESP.





### Data preparation

Make sure the data are prepared following the guidelines below:

- Prepare the dataset using Excel
- the first column of a spreadsheet always contains timestamps
- the first row of a spreadsheet always contains the name of variables
- The data for each variable are recorded in each corresponding column
- For the created spreadsheet, select the entire timestamp column, right-click on this column and go to "format cells", then go to "Time" or "Custom", change the format of time information to a format similar to: "mm/dd/yyyy h:mm" or "yyyy-mm-dd h:mm:ss".
- Save the dataset in .xlsx format

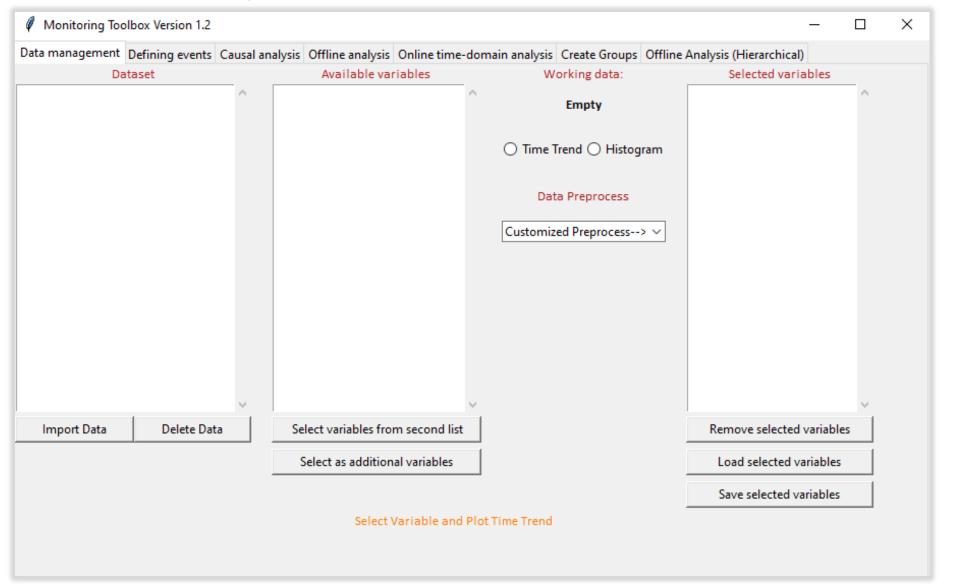
	А	В	С	D	E	F	G
1		Tubing temperature	Frequency	Motor Current	TC1	TC2	TC3
2	3/9/12 22:15	-1.364190189	0.675286424	0.49189377	-3.652629904	-3.478063792	-3.34500166
3	3/9/12 22:20	-1.498604965	0.675286424	0.587950416	-3.424377513	-3.2551307	-3.136641334
4	3/9/12 22:25	-1.302343733	0.675286424	0.49189377	-3.078778886	-2.977938815	-2.831244138
5	3/9/12 22:30	-1.184504911	0.675286424	0.49189377	-2.731496714	-2.494617748	-2.220262236
6	3/9/12 22:35	-1.118941918	0.675286424	0.324671085	-2.365448872	-2.020655317	-1.881266767
7	3/9/12 22:40	-1.053378925	0.675286424	0.49189377	-1.999401031	-1.690395652	-1.573328509
8	3/9/12 22:45	-0.987815931	0.675286424	0.65817481	-1.63335319	-1.556523086	-1.397895669
9	3/9/12 22:50	-0.894946846	0.675286424	0.49189377	-1.267305349	-1.474119634	-1.308935299
10	3/9/12 22:55	-0.67836475	0.675286424	0.230444806	-1.020548617	-1.359993647	-1.335386895
11	3/9/12 23:00	-0.594235033	0.675286424	0.322946051	-0.859315487	-1.276574446	-1.370401165
12	3/9/12 23:05	-0.738220634	0.675286424	0.230444806	-0.698082356	-1.269349947	-1.343913806
13	3/9/12 23:10	-0.550686357	0.675286424	0.230444806	-0.630755193	-1.262124479	-1.295956454
14	3/9/12 23:15	-0.552768548	0.675286424	0.230444806	-0.572068276	-1.254899011	-1.238107421
15	3/9/12 23:20	-0.554850738	0.675286424	0.49189377	-0.513381358	-1.230904492	-1.211772778
16	3/9/12 23:25	-0.527884322	0.675286424	0.401066418	-0.454694441	-1.202520661	-1.169750048



#### To start

Make sure the data sets to be used are in the same folder

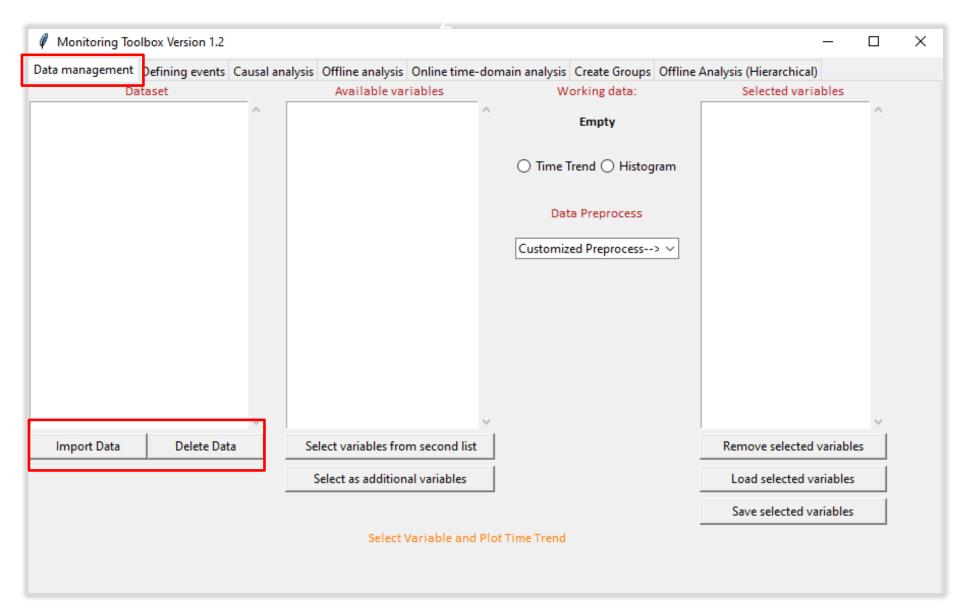
Click "Monitoring\_v1.2.exe" to start the program

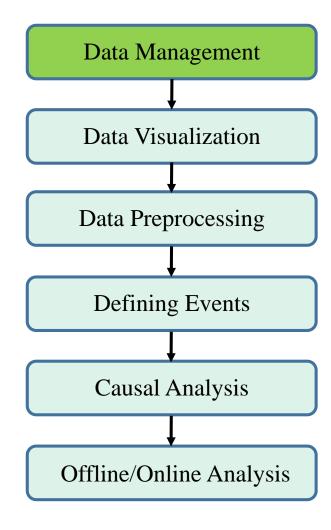




### Data management – Import data sets

Make sure the data sets to be used are in the same folder

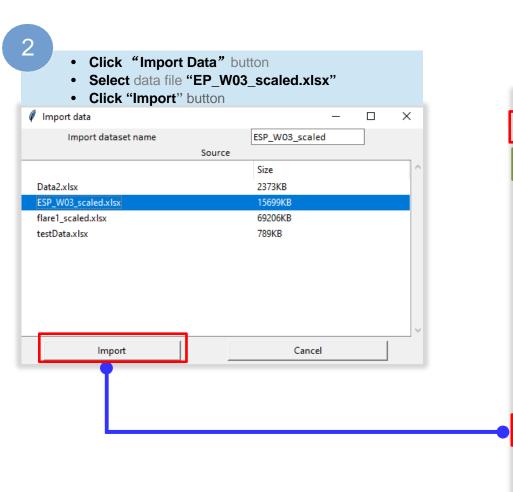


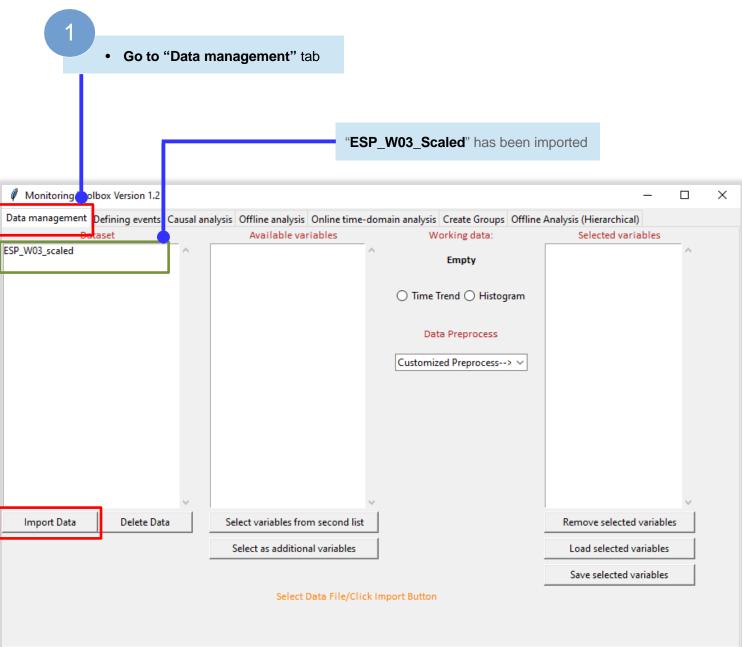




### Data management – Import data sets

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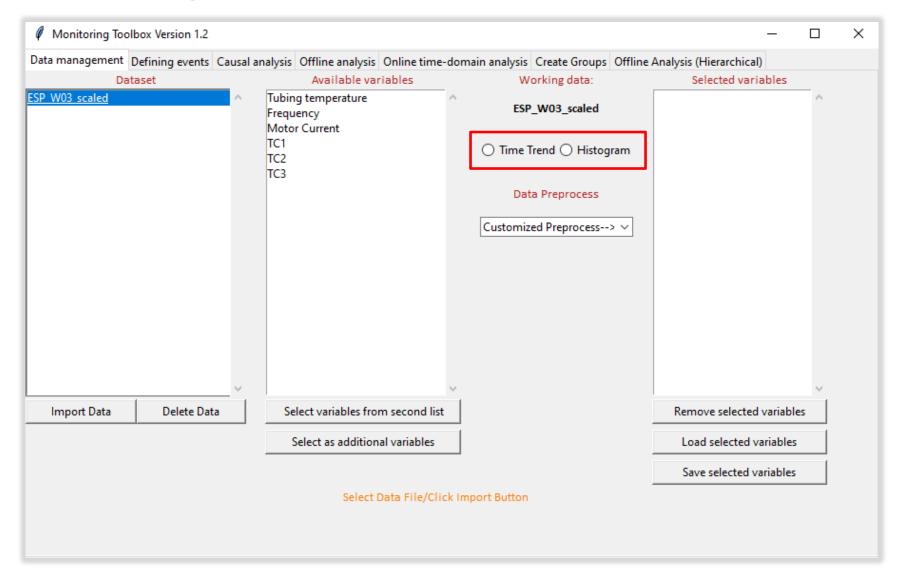


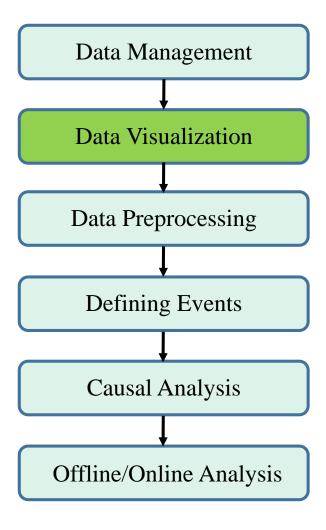
Note: Please save data sets in the ".xlsx" format



#### Data visualization

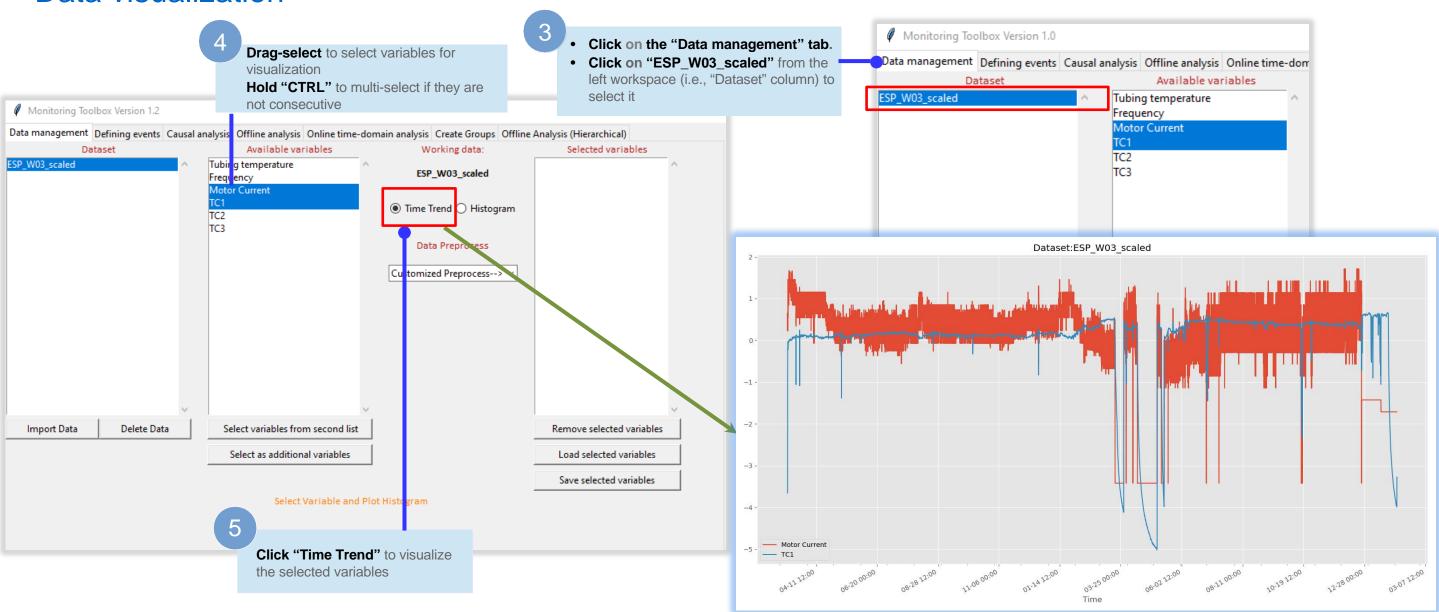
Under the "Data management" tab, the variables can be visualized





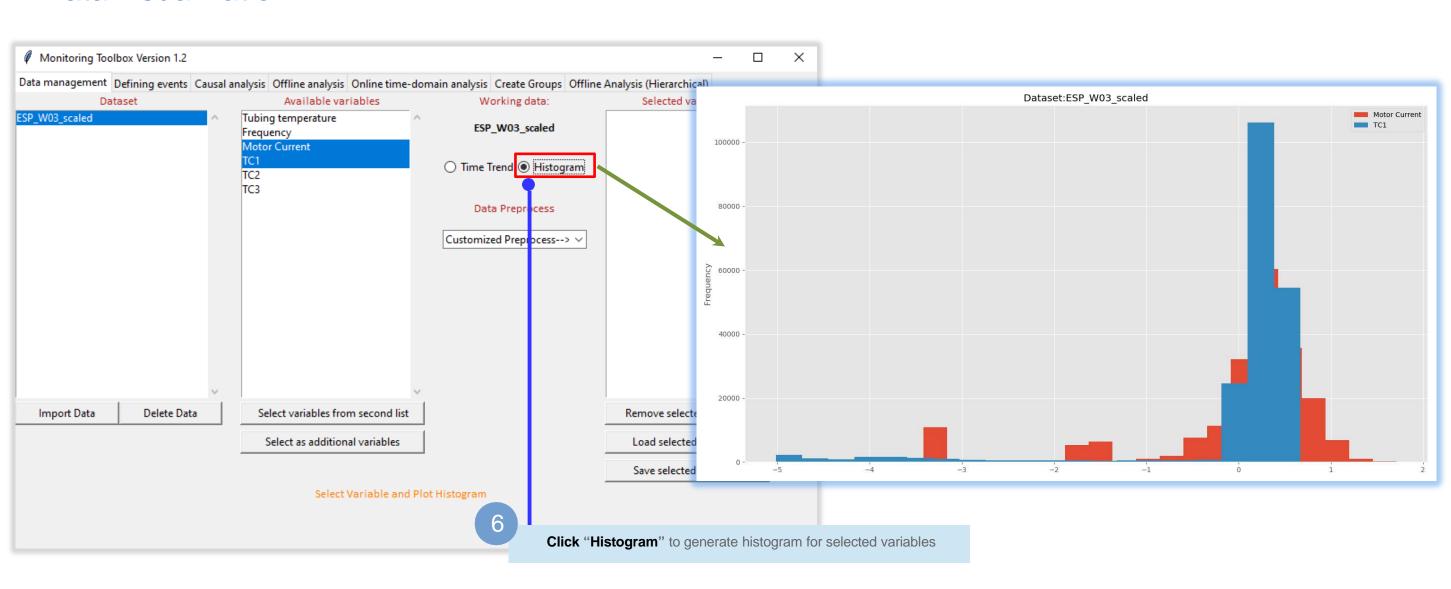


#### Data visualization





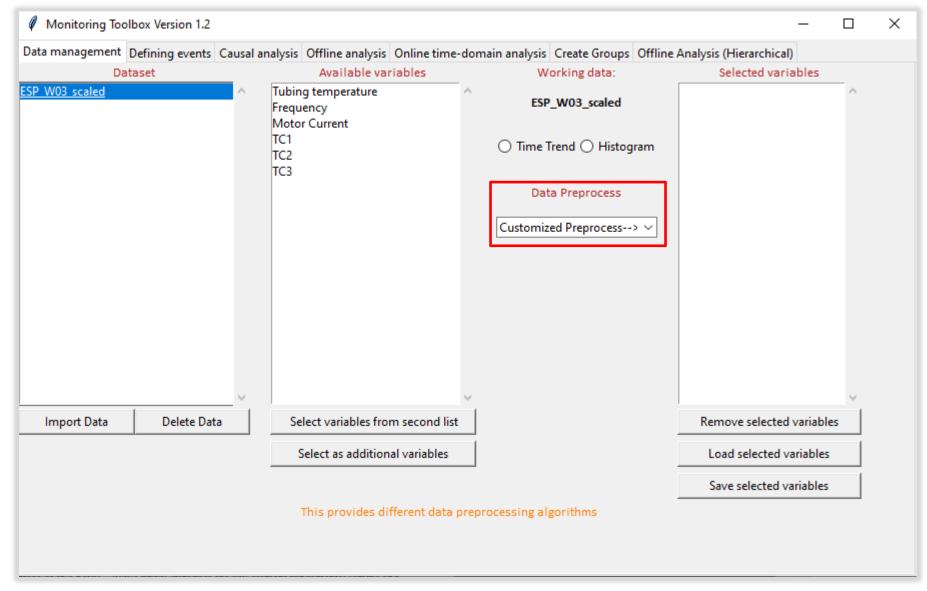
#### Data visualization

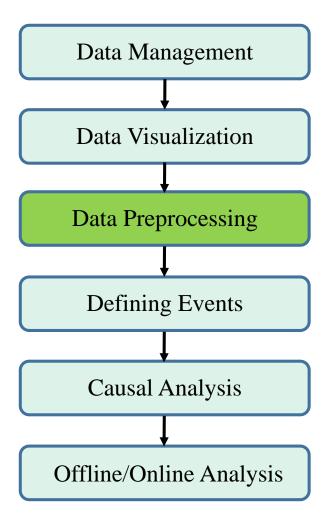




### Data preprocessing

Under the "Data management" tab, the data can be preprocessed before further analysis is conducted



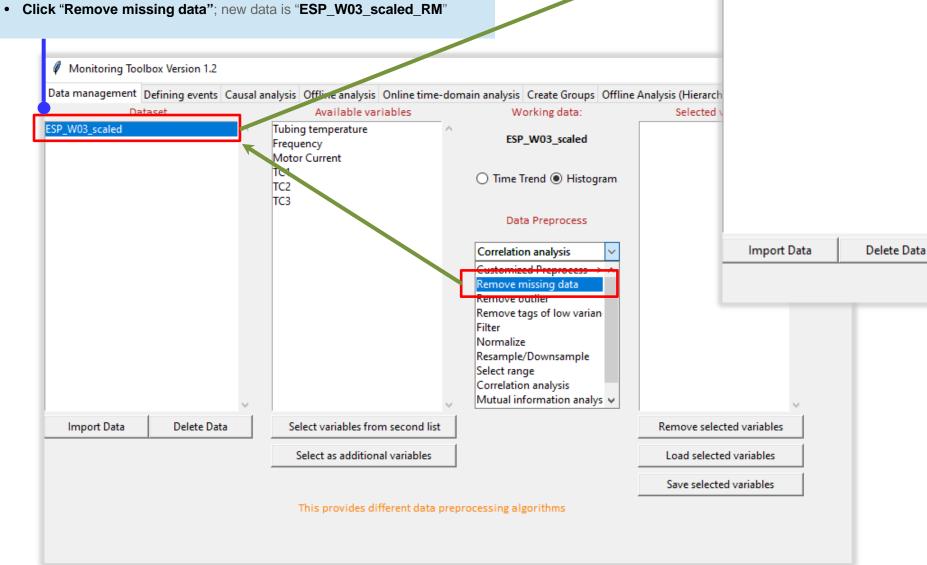




This step is to remove missing data from the dataset

• Select "ESP\_W03\_scaled" from Dataset column

• Click "Customized Preprocess -->" on the right





Available variables

Select variables from second list

Select as additional variables

Tubing temperature

Frequency Motor Current

TC1

TC2

TC3

Monitoring Toolbox Version 1.2

ESP W03 scaled

ESP\_W03\_scaled\_RM

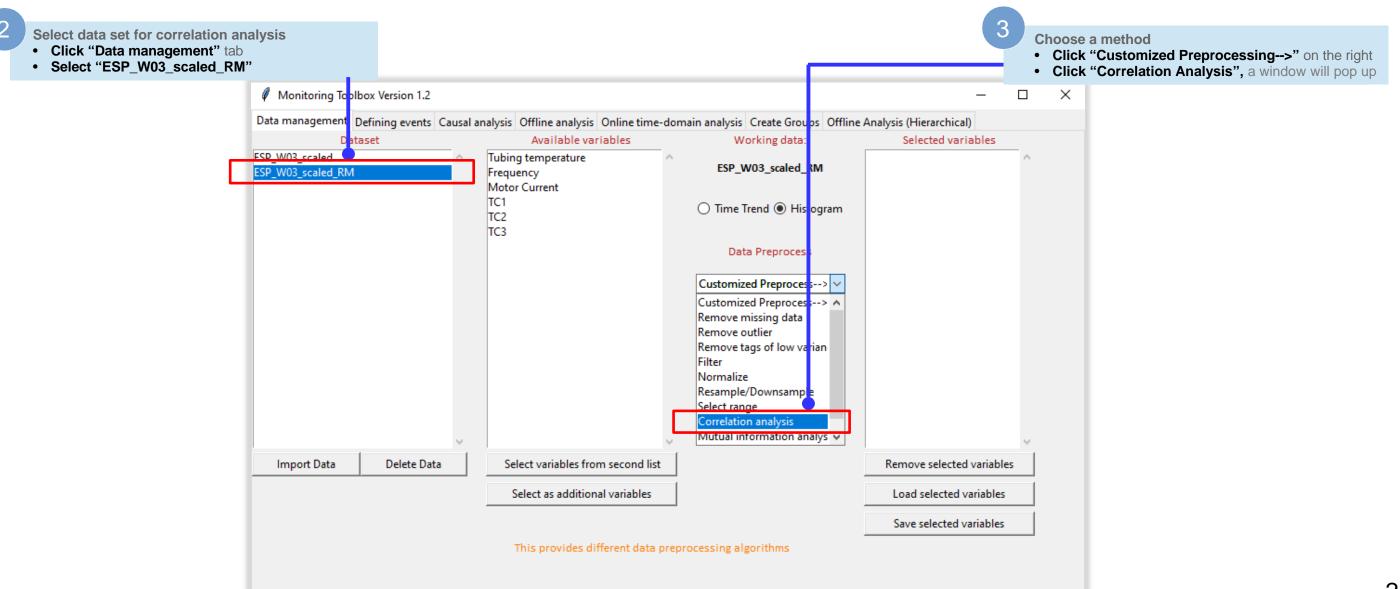
Dataset

Data management Defining events Causal analysis Offline analysis Online time-dom



### Data preprocessing – Correlation analysis

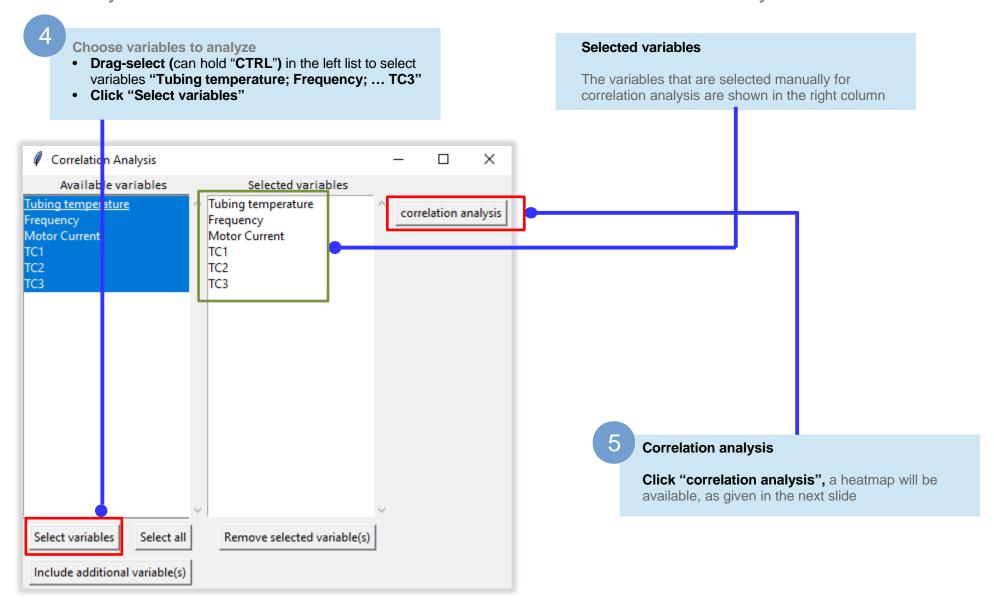
Conduct correlation analysis to check the influential variables as well as the co-linearity of the data-set.





### Data preprocessing – Correlation analysis

Conduct correlation analysis to check the influential variables as well as the co-linearity of the data-set.





### Correlation analysis – Result

#### Results

Variables are highly correlated to each other if the absolute value of the corresponding index is large.

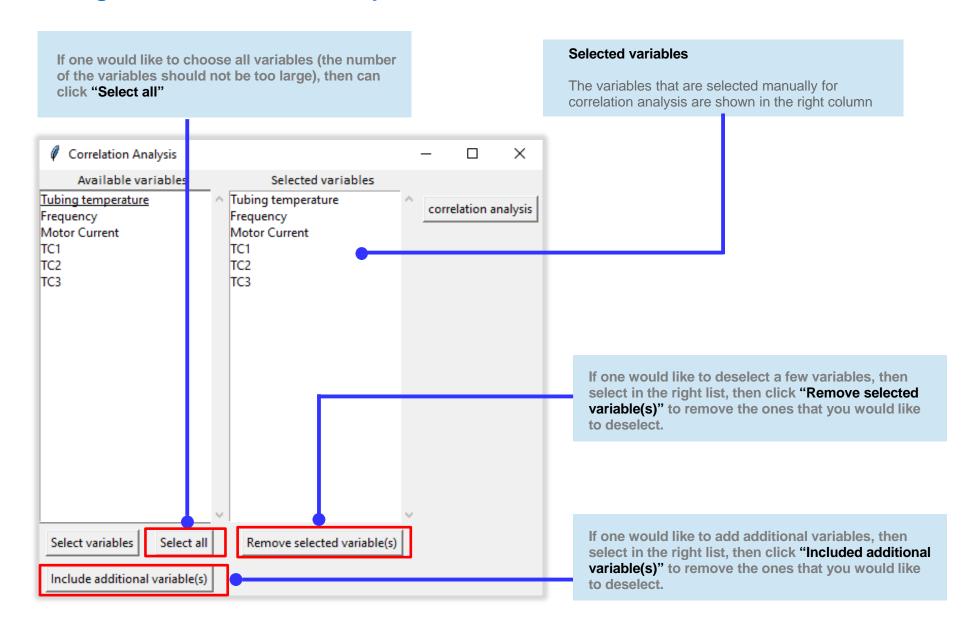
One can select more variables and get more information from the correlation test results.

5 702  $\overline{C}$ 1 0.76 0.7 0.95 0.93 0.93 Tubing temperature 0.96 0.76 0.85 0.8 0.76 0.76 Frequency 0.90 0.7 0.85 0.75 0.64 0.64 Motor Current 0.84 0.95 0.8 0.75 0.96 TC1 0.96 0.78 0.93 0.76 0.64 0.96 TC2 0.72 0.76 0.64 0.96 0.93 TC3 0.66

Note: Please close the current popup figures before taking next action

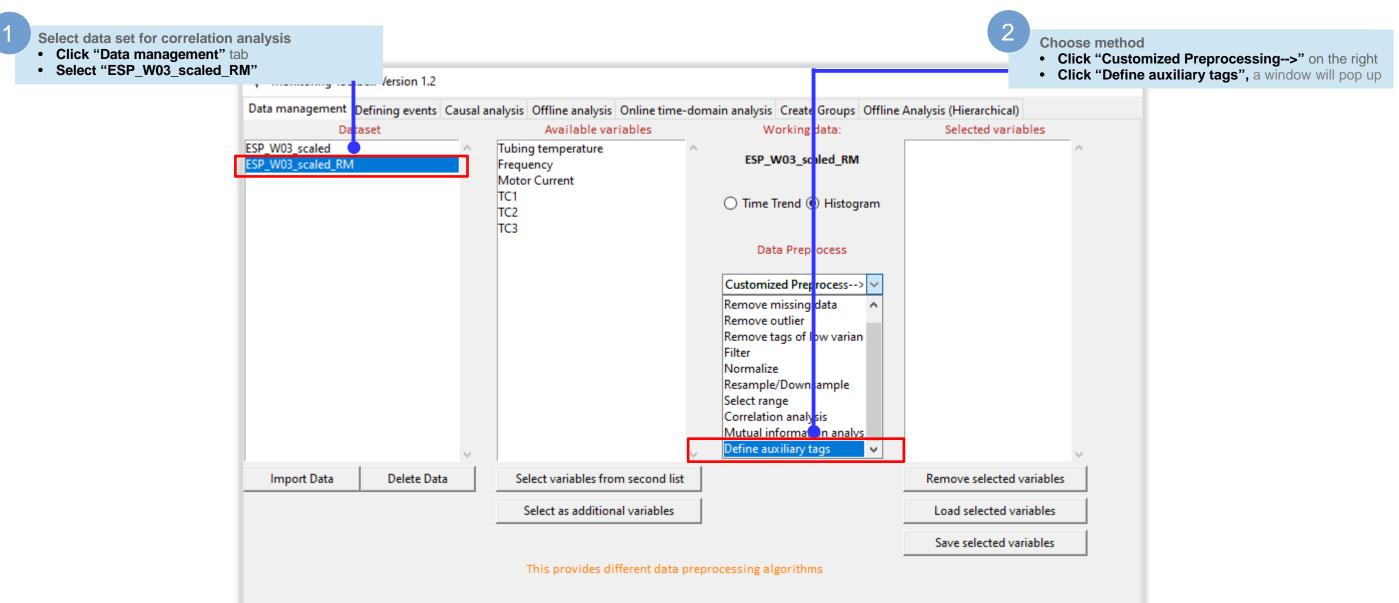


### Data preprocessing – Correlation analysis





### Data preprocessing – Defining auxiliary tags

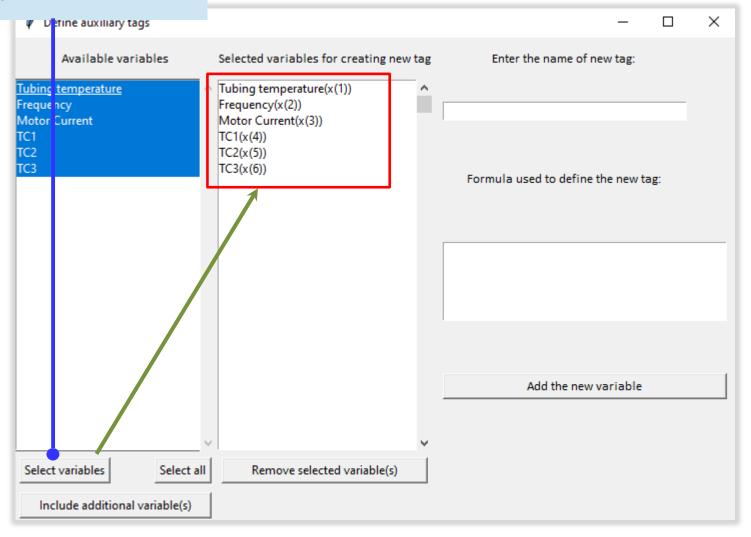




### Data preprocessing – Defining auxiliary tags

Choose variables to analyze

- **Drag-select** (can hold "CTRL") in the left list to select variables "Tubing temperature; Frequency; ... TC3"
- Click "Select variables"





### Data preprocessing – Defining auxiliary tags

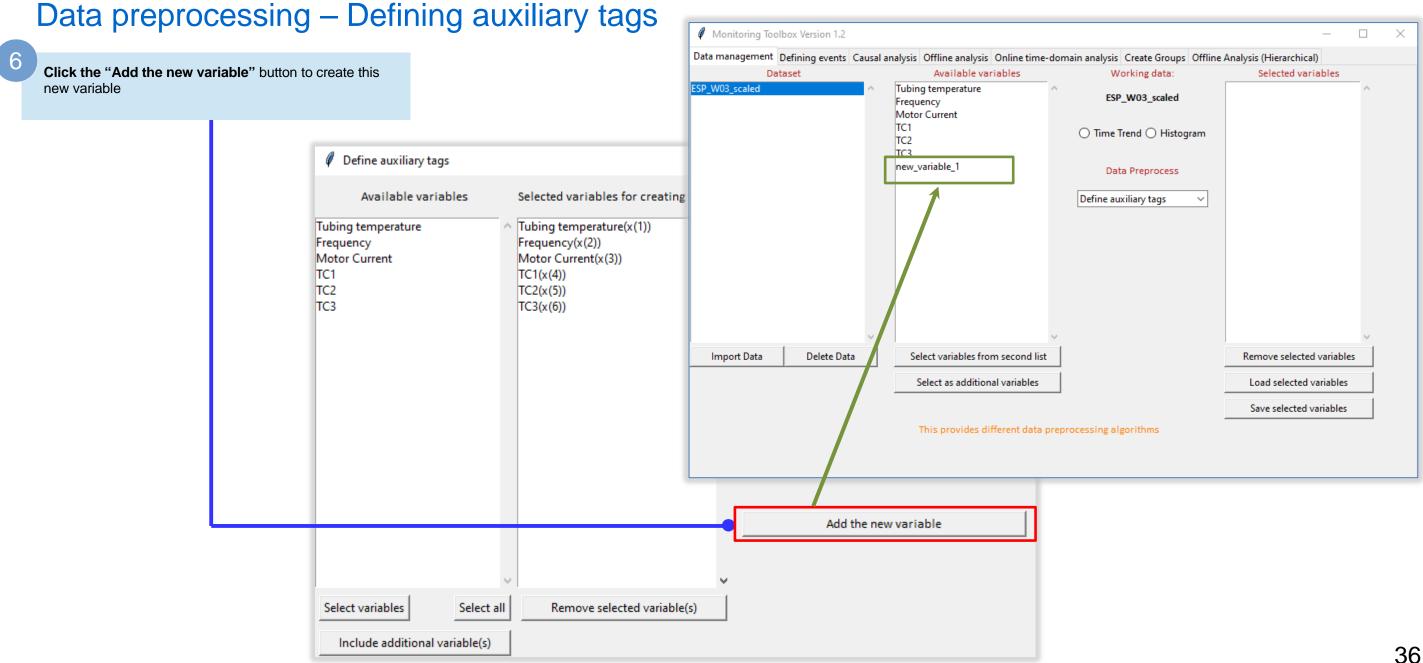
Define the name of the new variable • Enter "new variable 1" as the name of the newly defined variable Define auxiliary tags × Available variables Selected variables for creating new tag Enter the name of new tag: Tubing temperature(x(1)) Tubing temperature Frequency Frequency(x(2)) new\_variable\_1 IVIOTOr Current(X(3)) iviotor Current TC1 TC1(x(4)) TC2 TC2(x(5)) TC3(x(6)) TC3 Formula used to define the new tag: Add the new variable Select variables Remove selected variable(s) Select all Include additional variable(s)



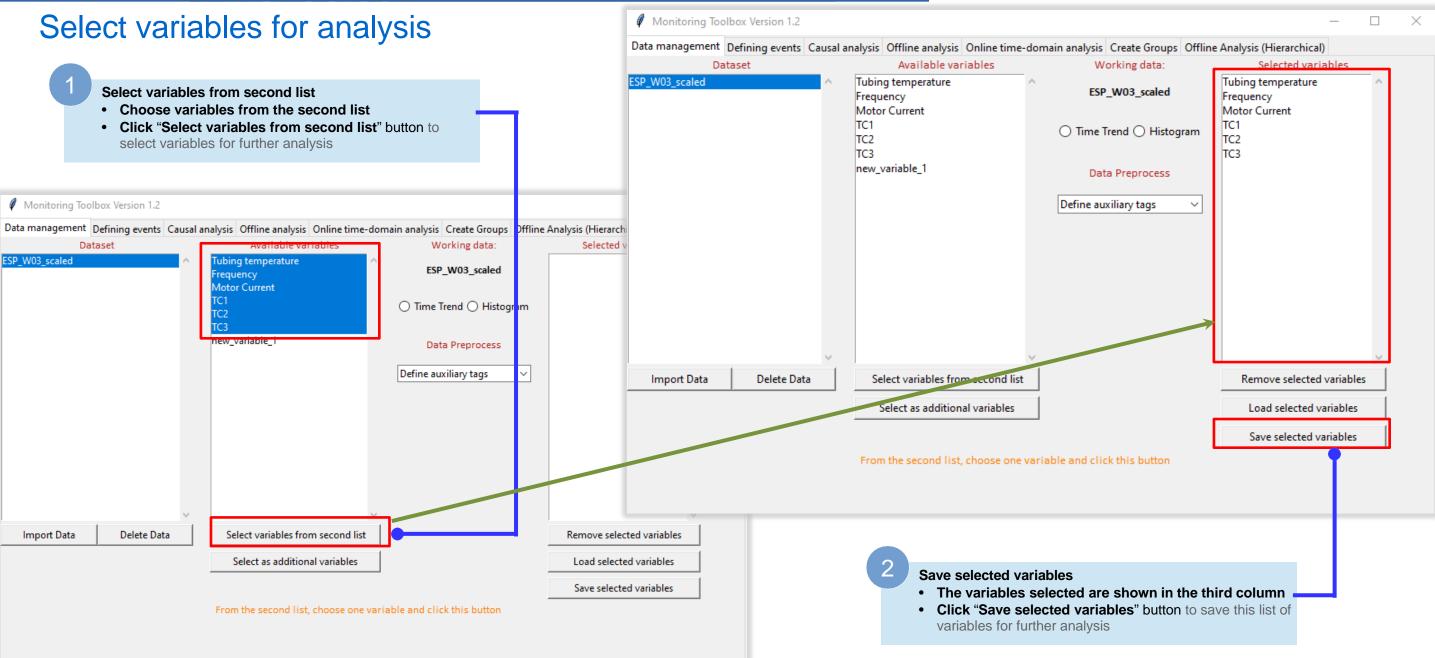
### Data preprocessing – Defining auxiliary tags

5 Enter the formula used for defining the new variable • Enter "x(4)/x(5)" - the new variable is calculated in a way that TC1 is divided by TC2 Define auxiliary tags × Available variables Selected variables for creating new tag Enter the name of new tag: Tubing temperature(x(1)) Tubing temperature Frequency(x(2)) Frequency new\_variable\_1 Motor Current(x(3)) Motor Current TC1 TC1(x(4)) TC2 TC2(x(5)) TC3(x(6)) TC3 Formula used to define the new tag: x(4)/x(5) Add the new variable Select variables Remove selected variable(s) Select all Include additional variable(s)





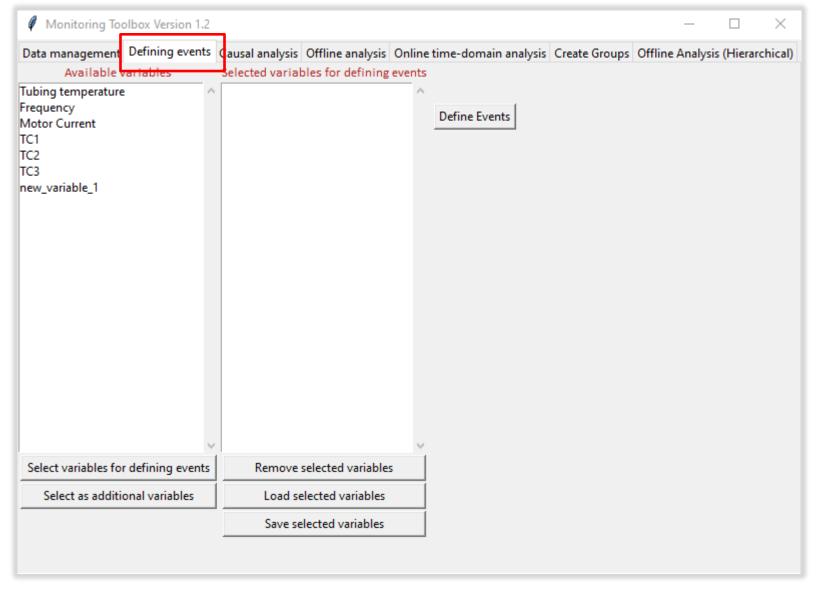


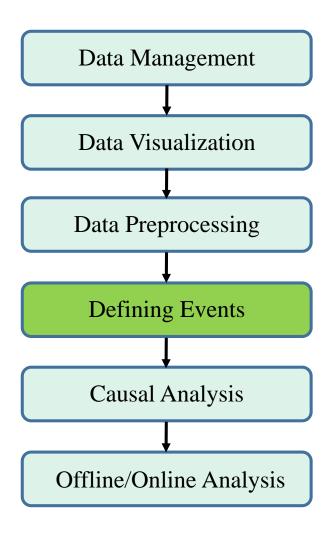




#### **Defining Events**

Under the "Defining events" tab, different types of events that took place can be defined

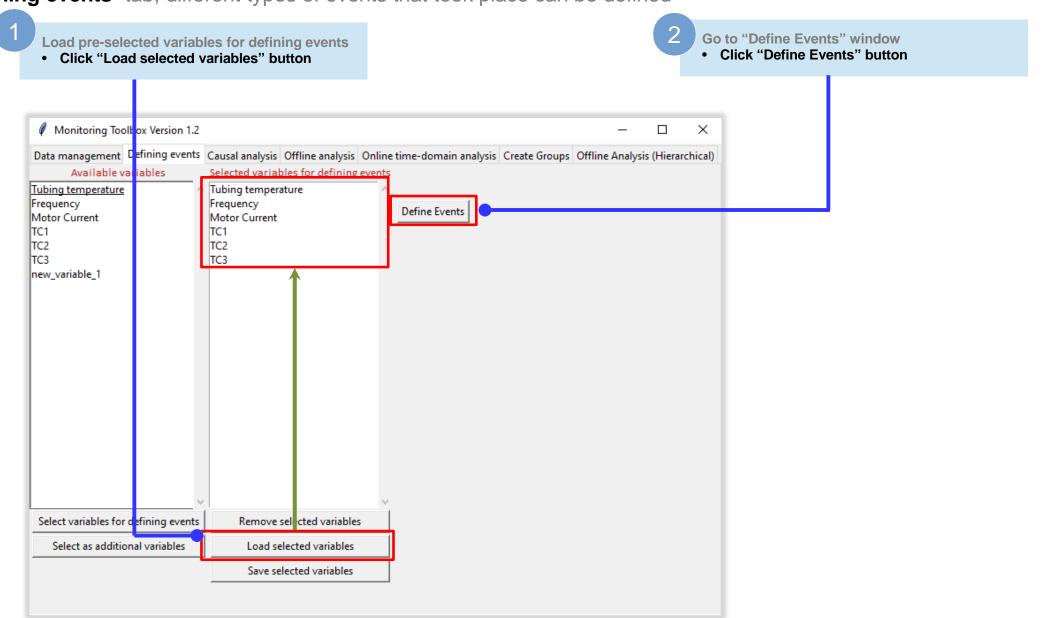






#### **Defining Events**

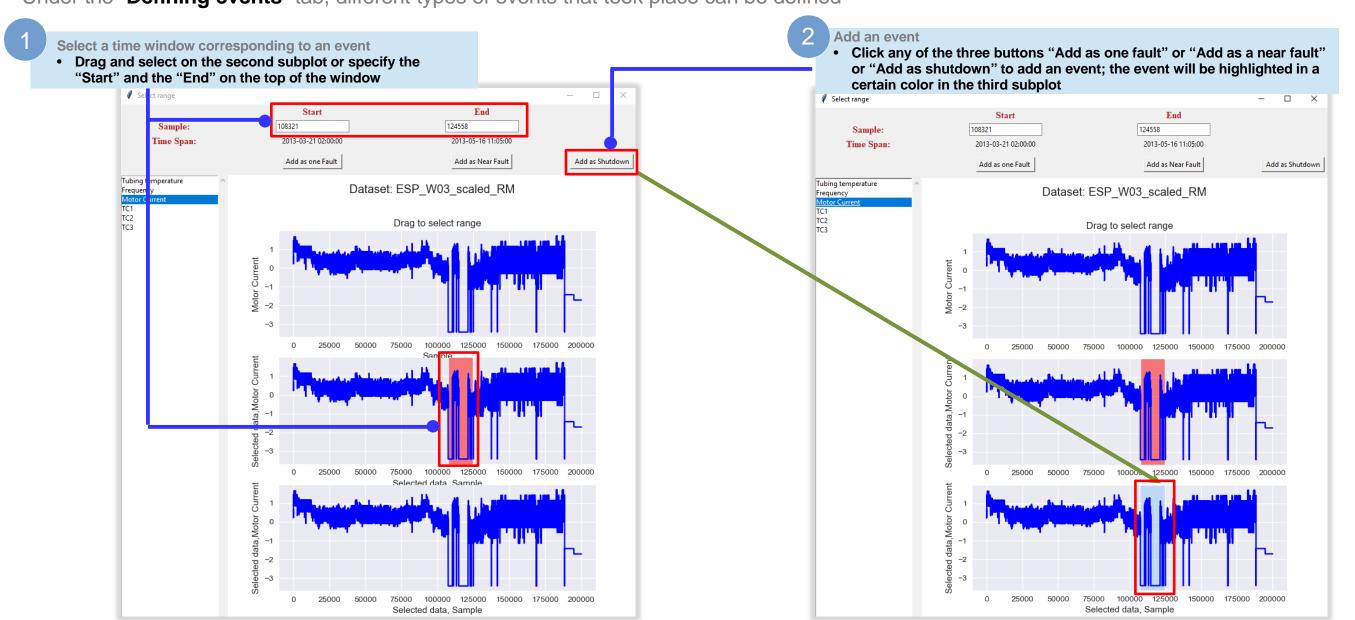
Under the "Defining events" tab, different types of events that took place can be defined





#### **Defining Events**

Under the "Defining events" tab, different types of events that took place can be defined



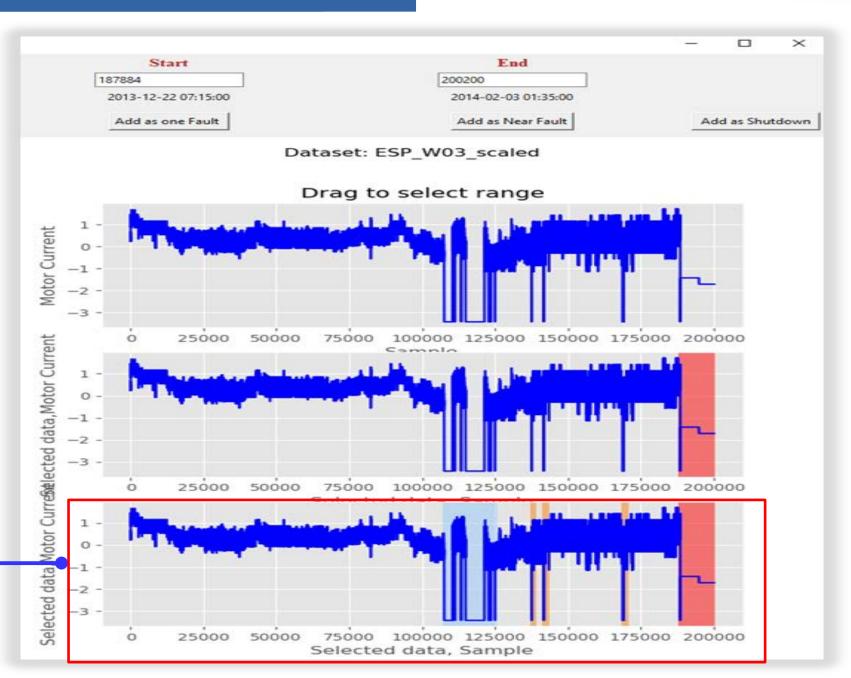


#### **Defining Events**

Three types of events have been added

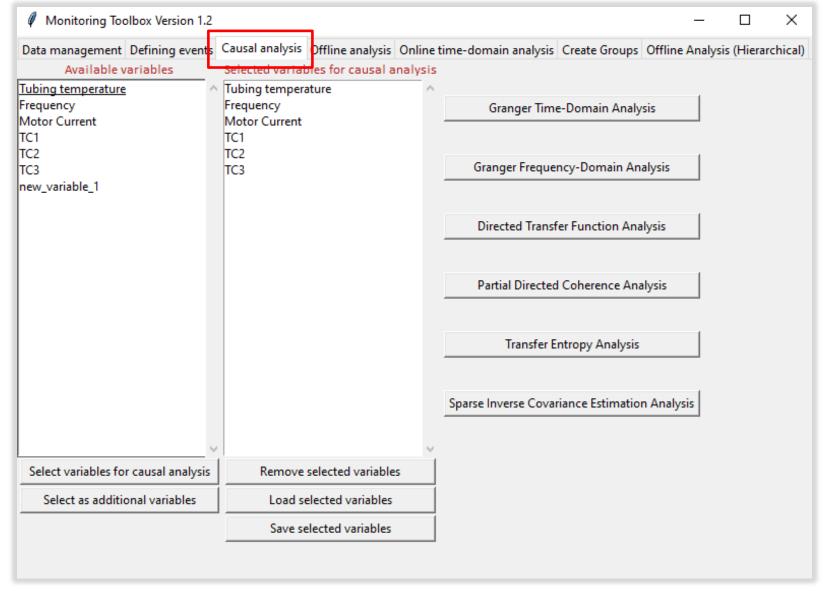
#### Select a time window corresponding to an event

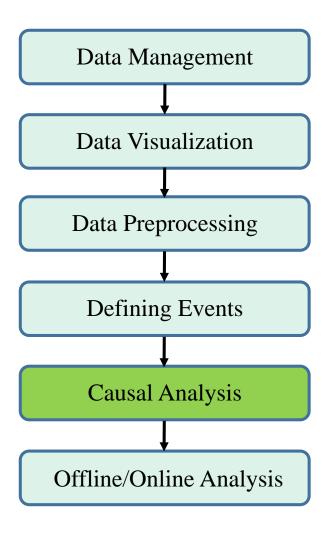
 Shutdown is highlighted in blue; near faults are highlighted in yellow; faults are highlighted in red in the third subplot





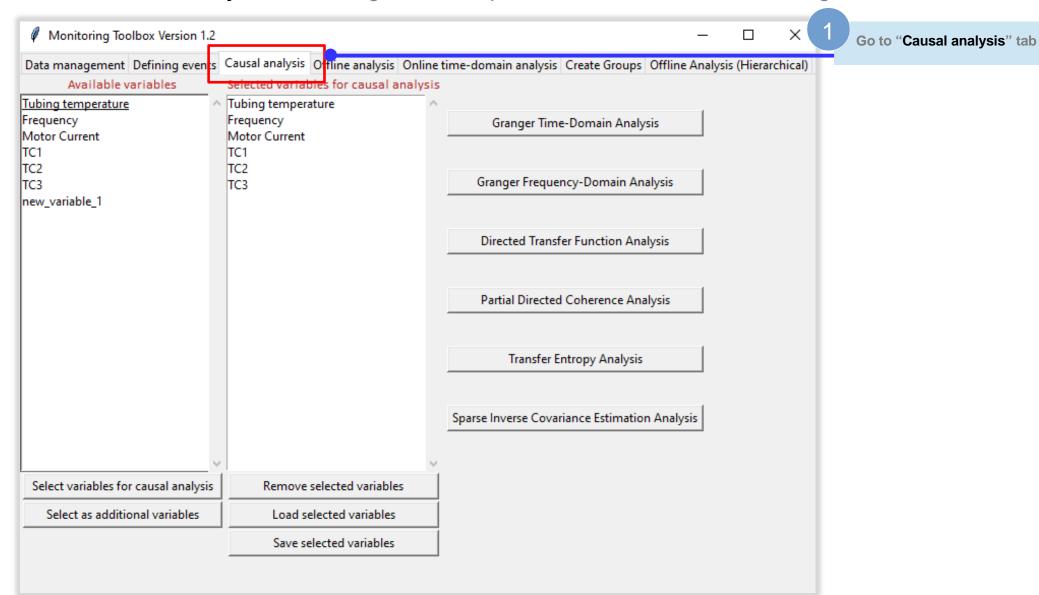
#### Causal analysis





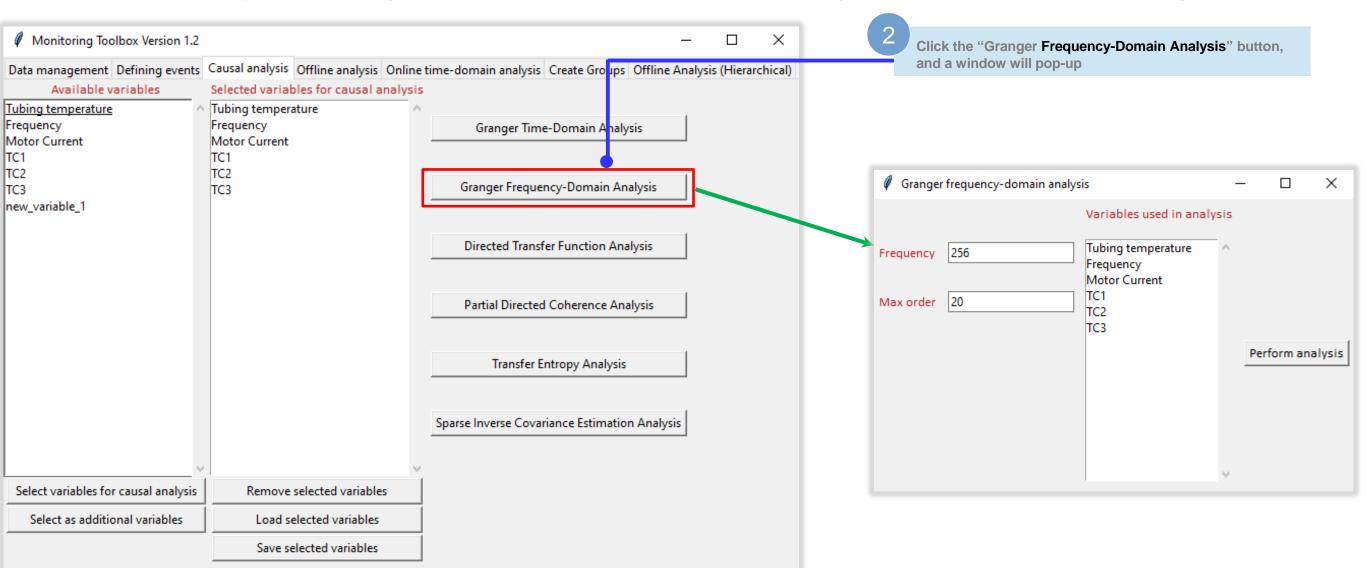


#### Causal analysis



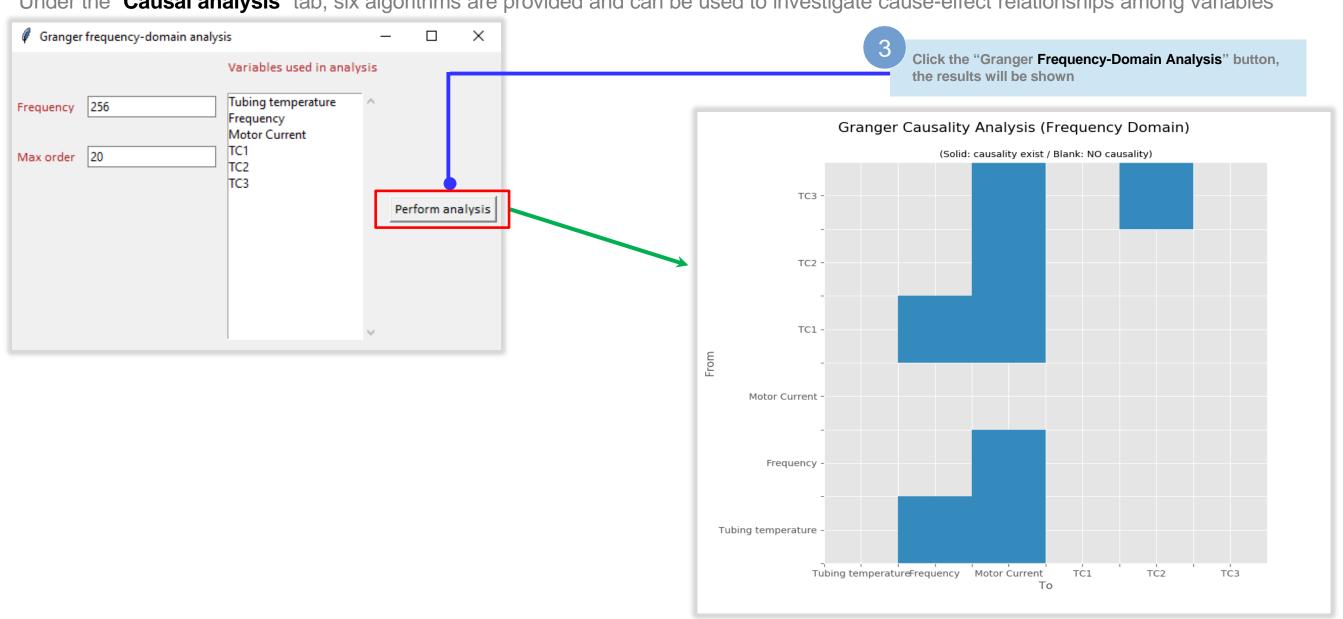


#### Causal analysis



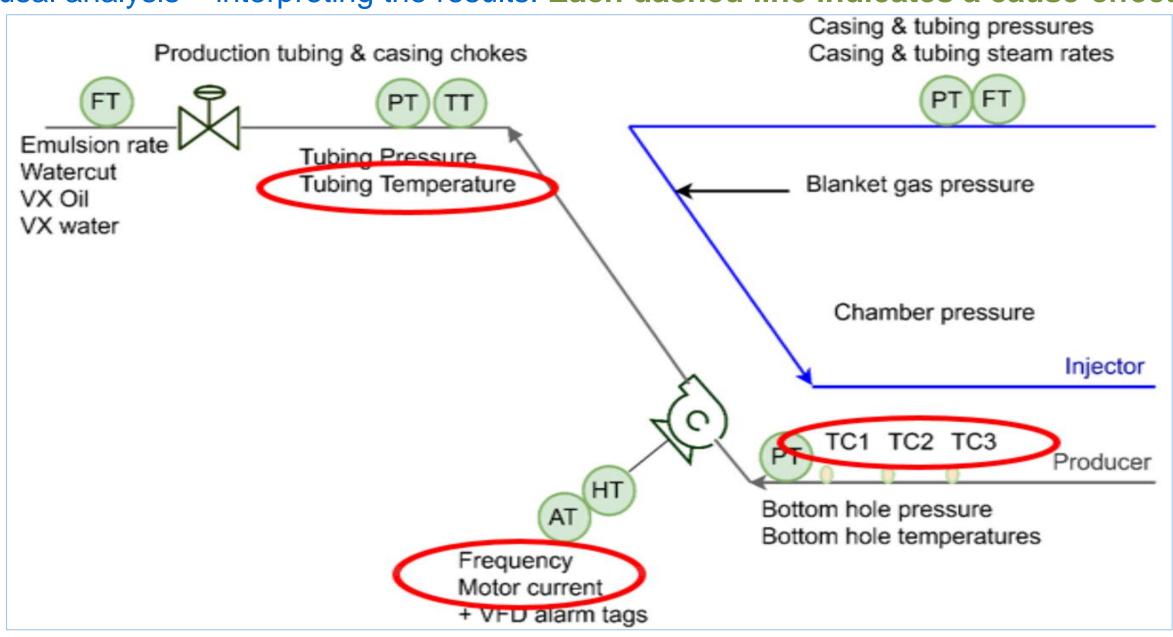


#### Causal analysis



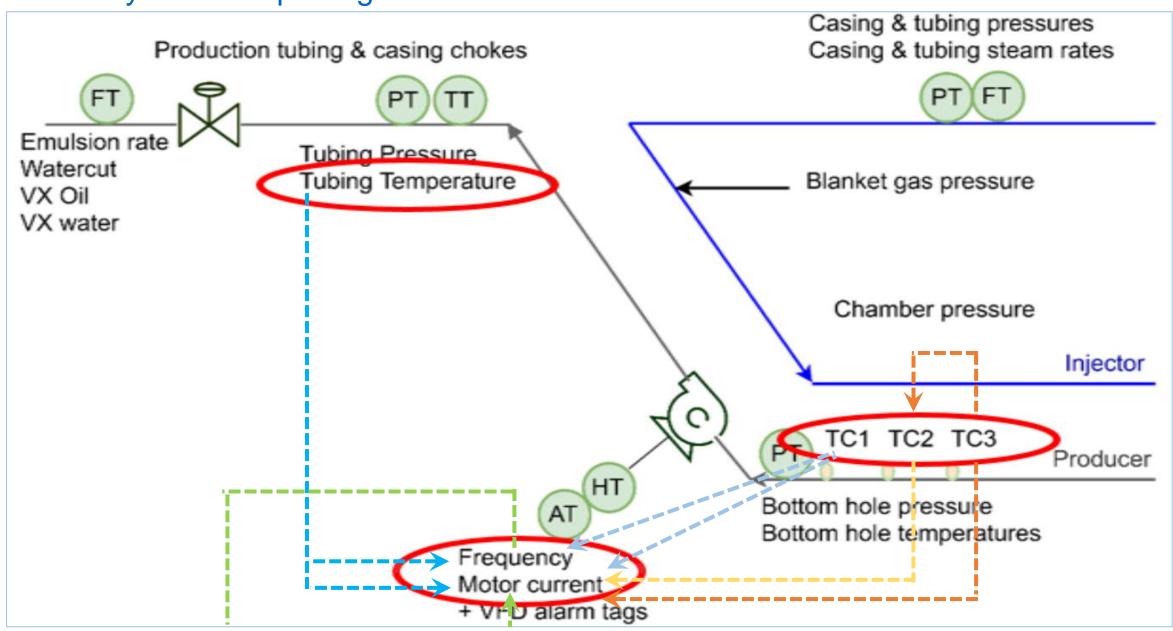


Causal analysis – interpreting the results: Each dashed line indicates a cause-effect relationship





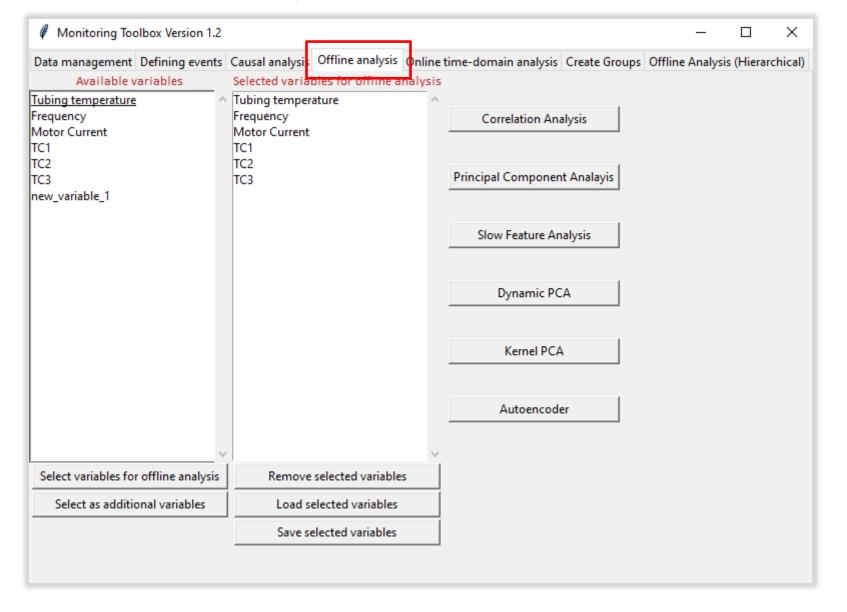
Causal analysis – interpreting the results: Each dashed line indicates a cause-effect relationship

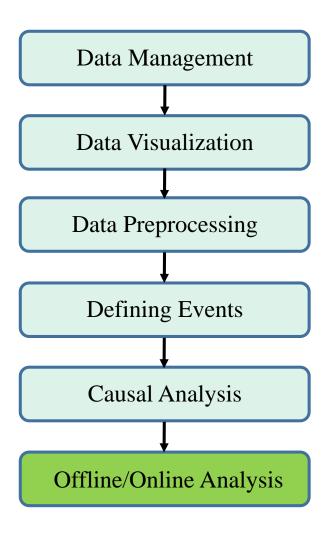




#### Offline monitoring

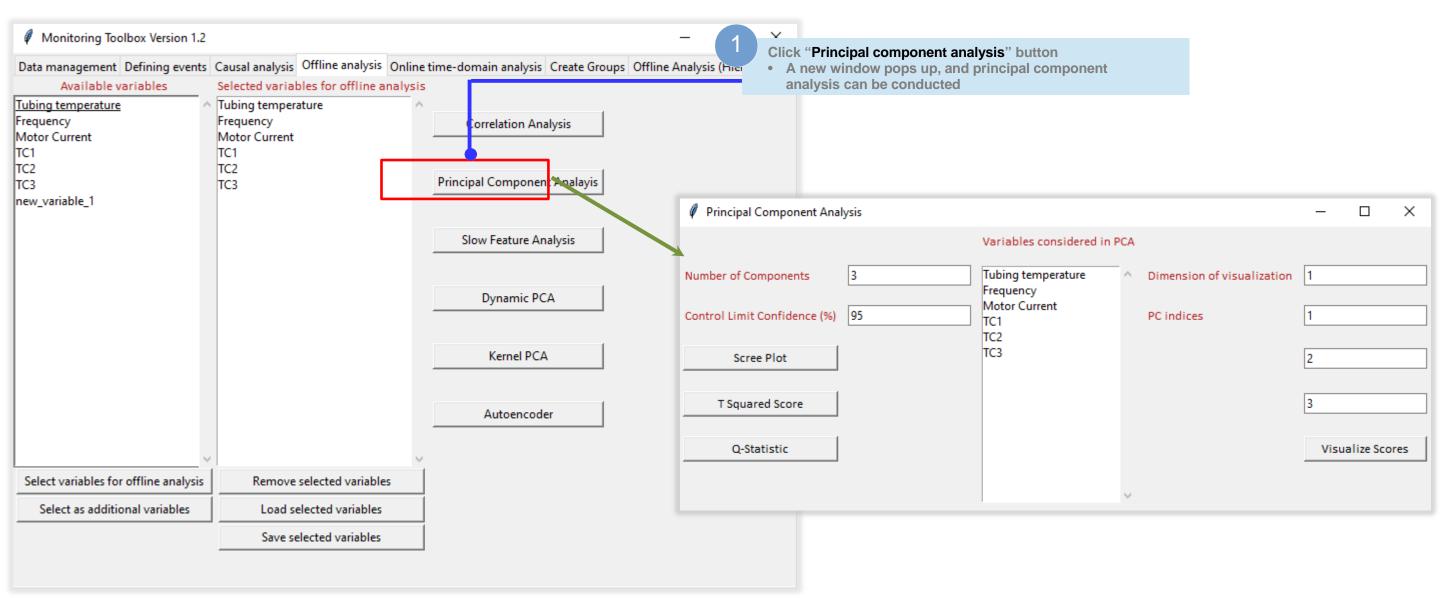
Under the "Offline analysis" tab, five algorithms are provided and can be used to monitor the process and detect events





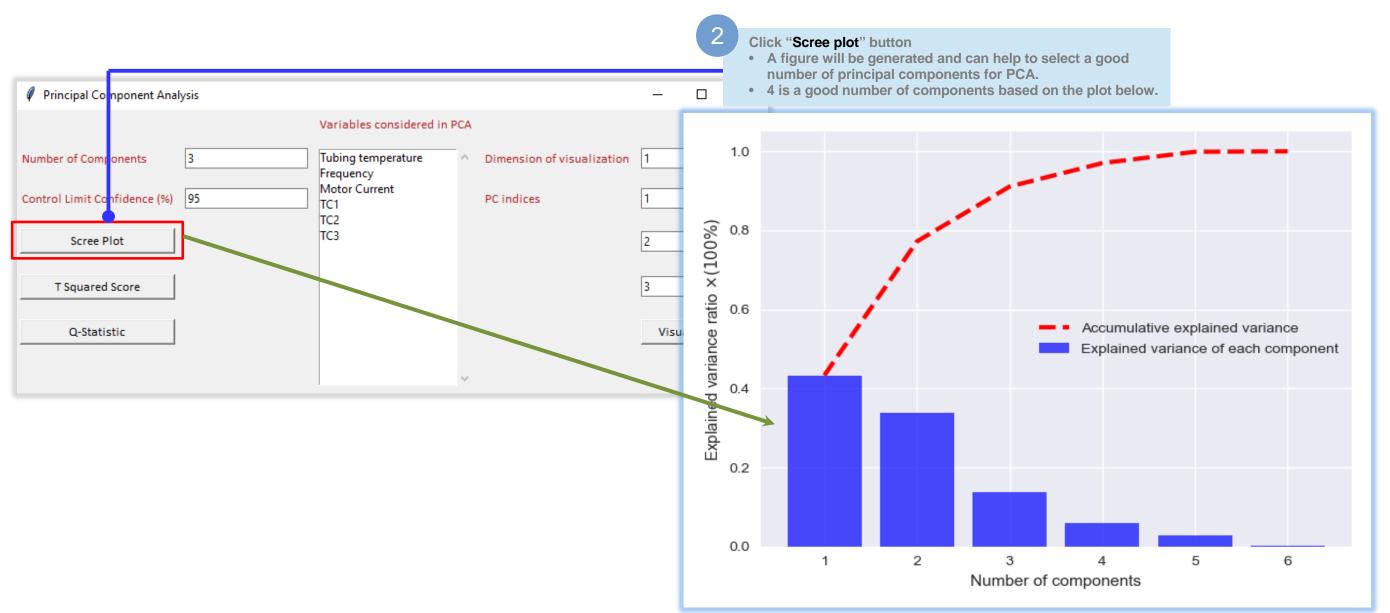


#### Offline monitoring – Principal component analysis





Offline monitoring – Principal component analysis (PCA)



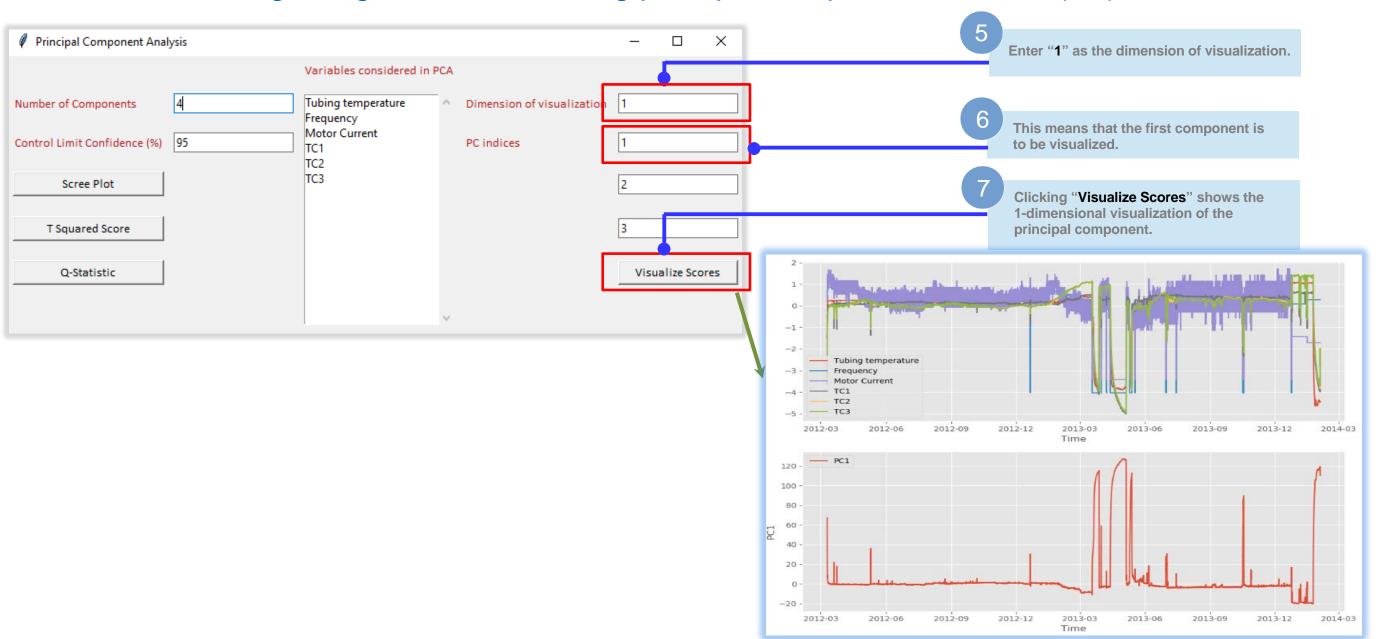


Offline monitoring – Principal component analysis (PCA)



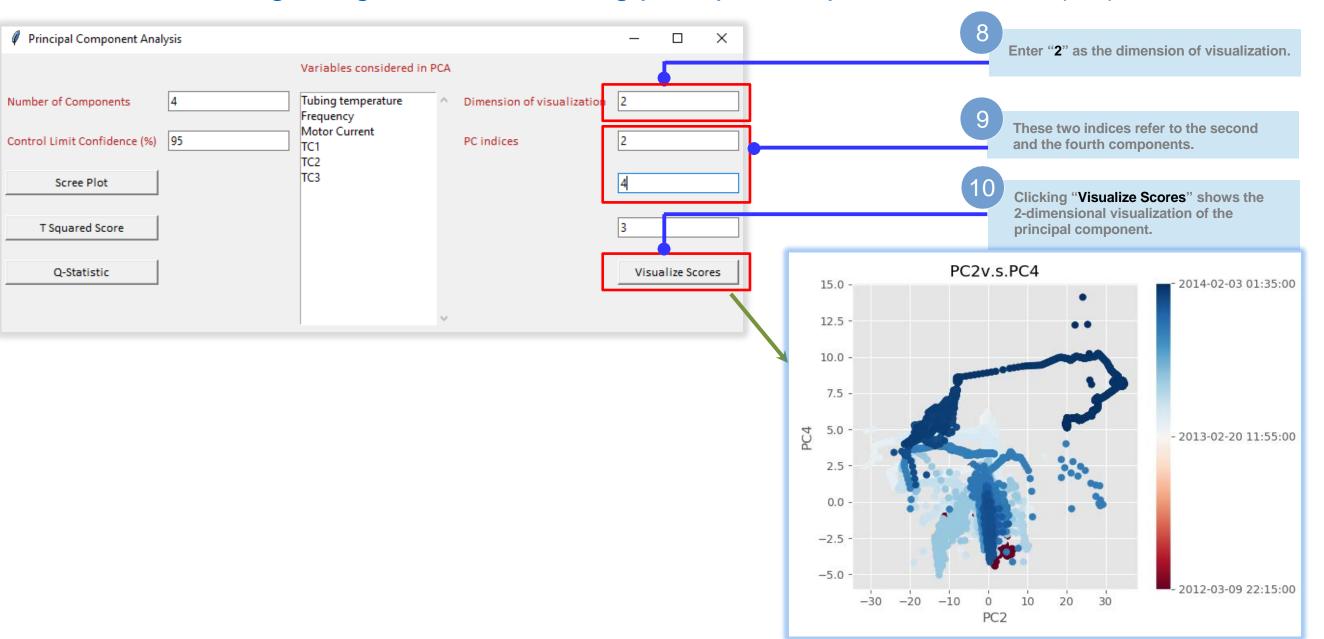


Offline monitoring using PCA – Visualizing principal components of PCA (1D)



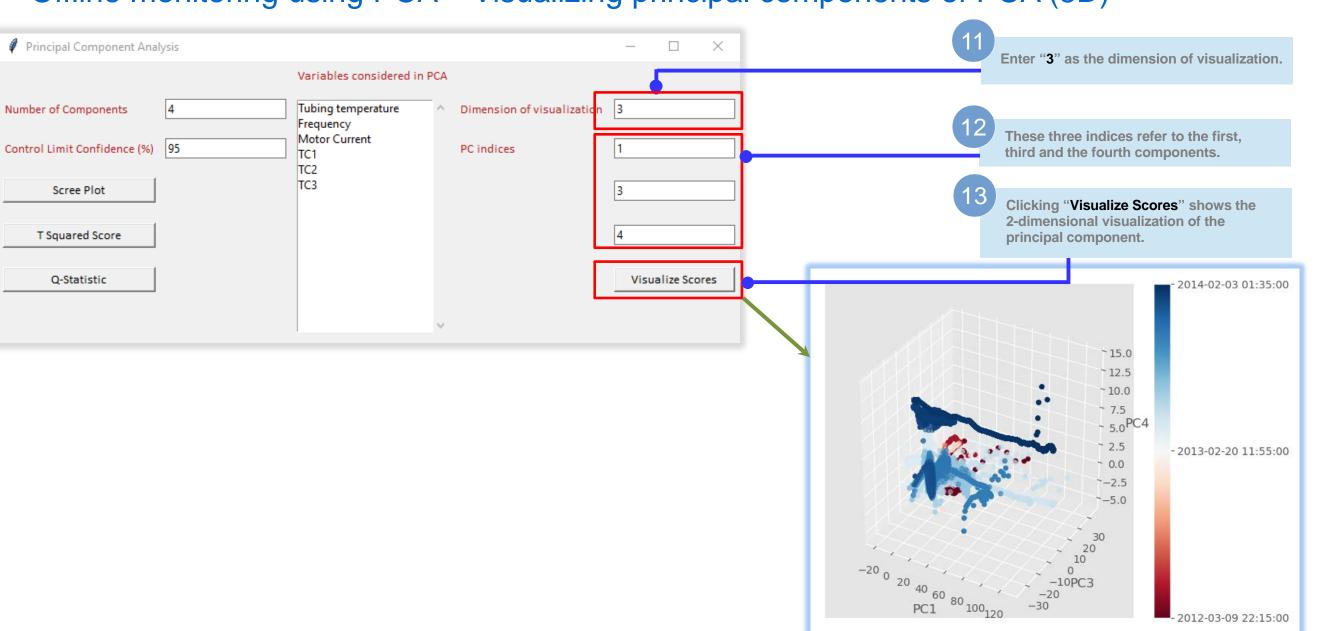


#### Offline monitoring using PCA – Visualizing principal components of PCA (2D)



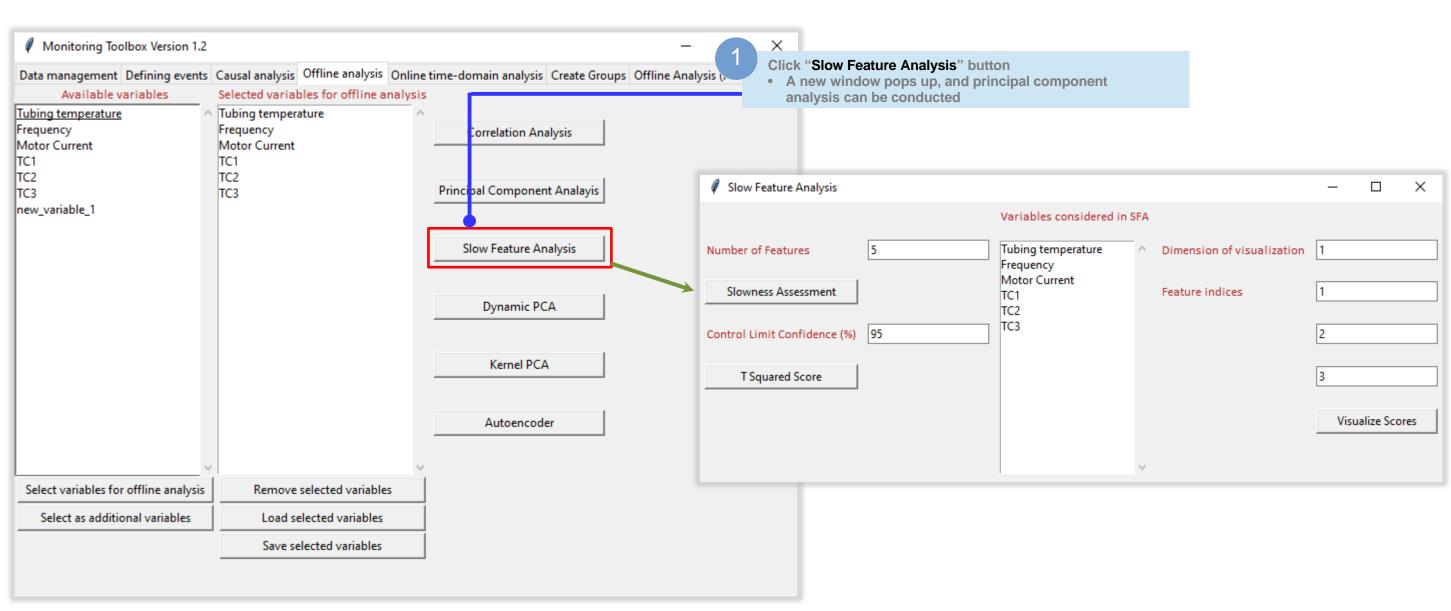


Offline monitoring using PCA – Visualizing principal components of PCA (3D)



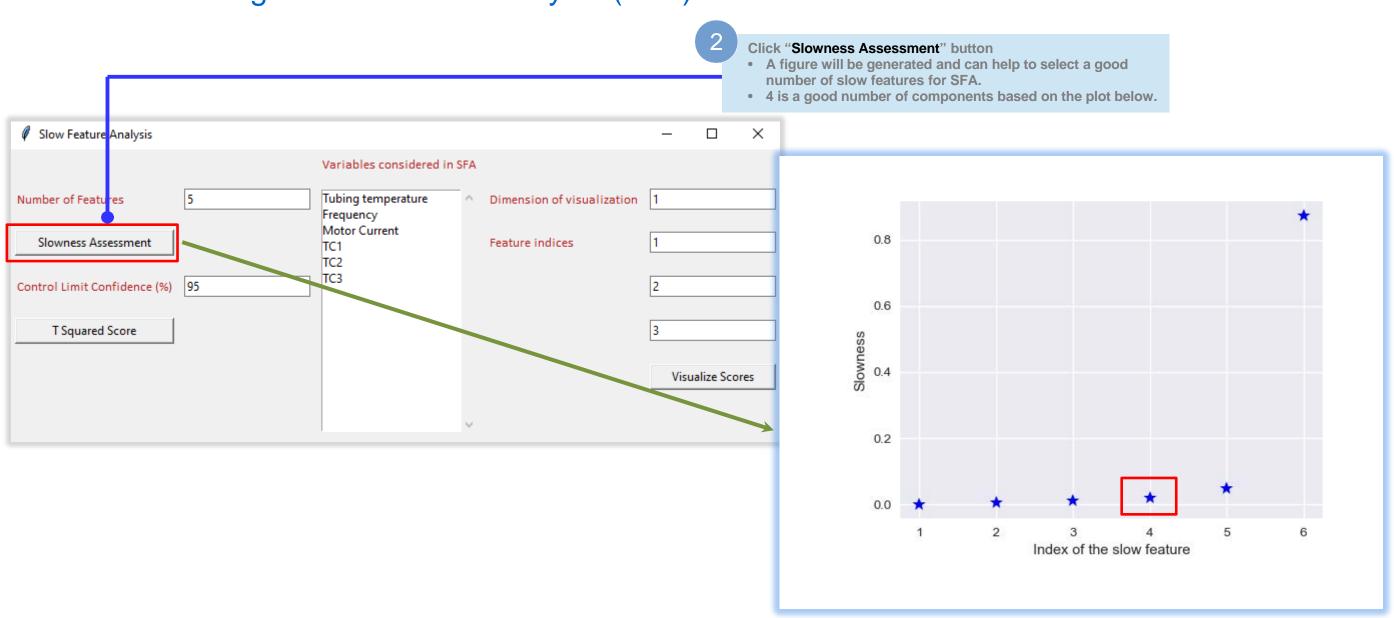


#### Offline monitoring – Slow feature analysis



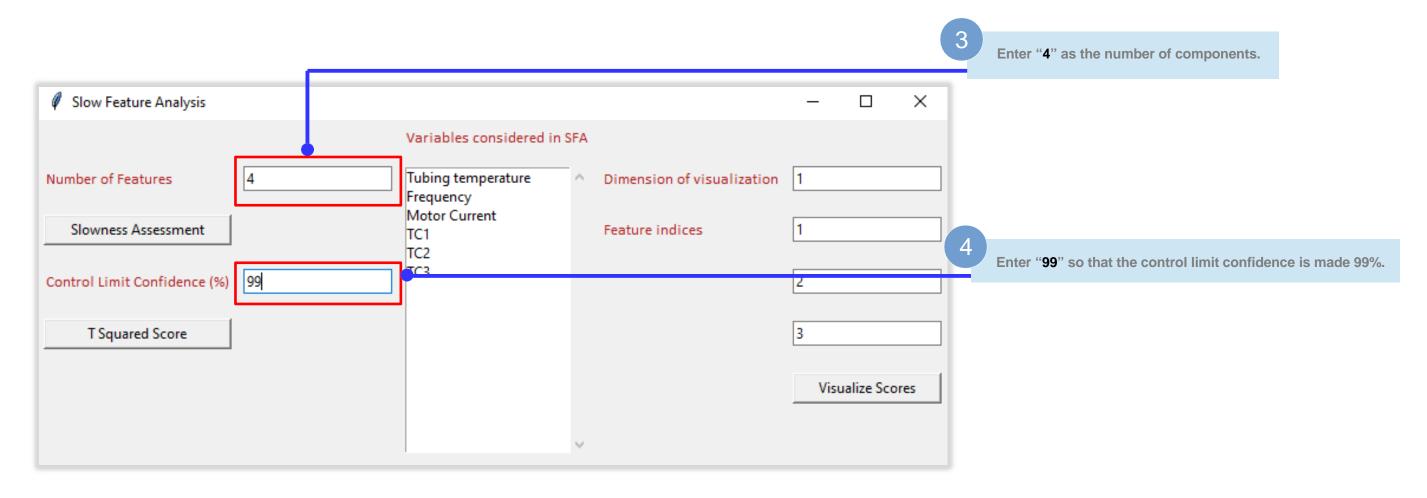


#### Offline monitoring – Slow feature analysis (SFA)



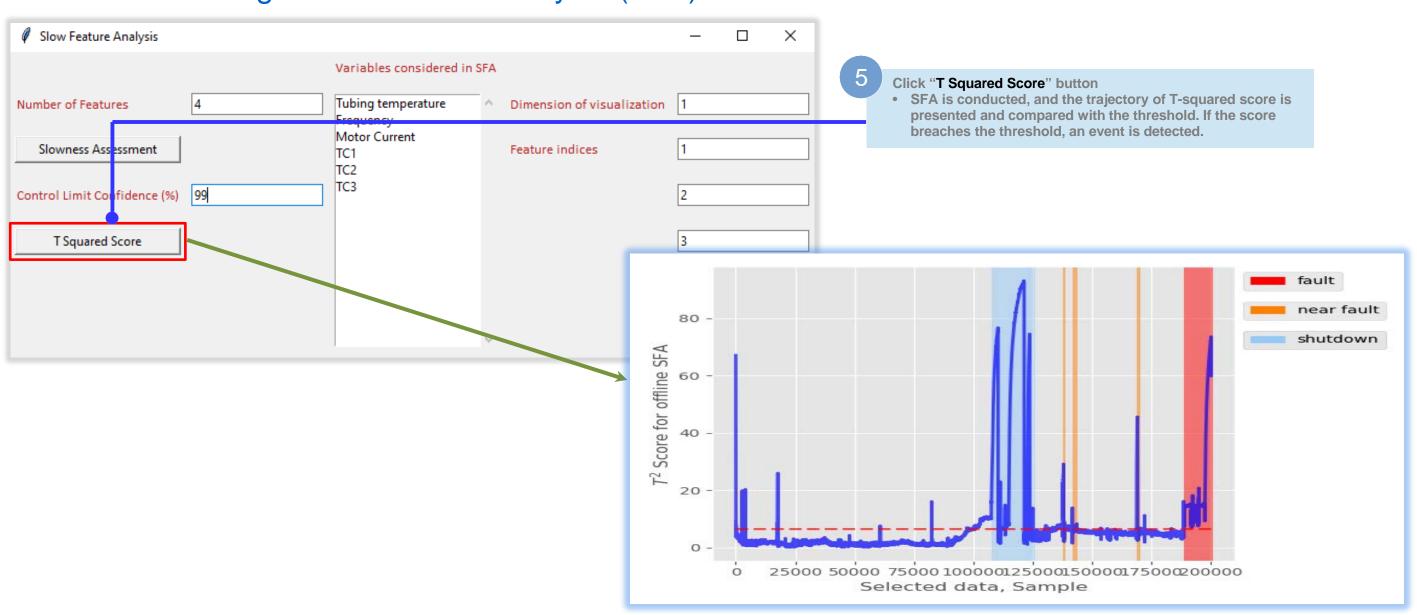


Offline monitoring – Slow feature analysis (SFA)





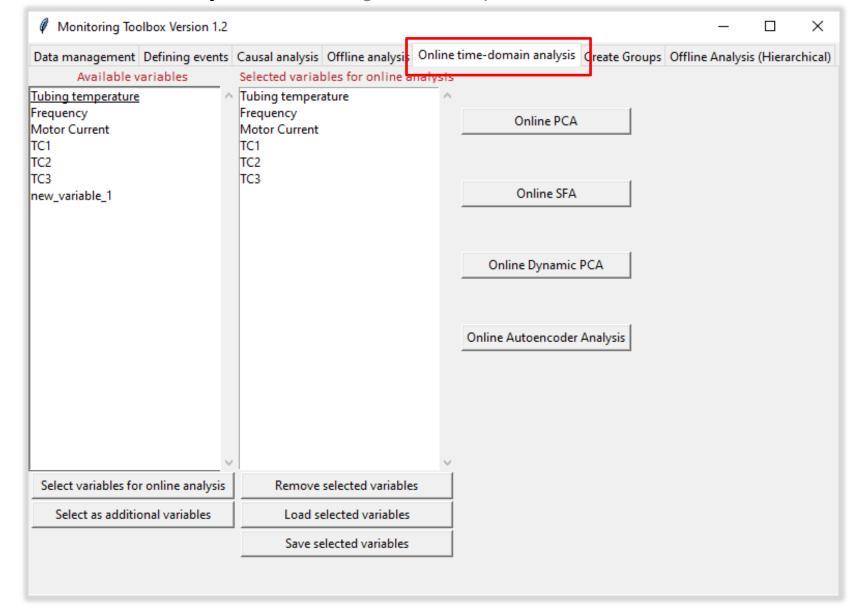
#### Offline monitoring – Slow feature analysis (SFA)

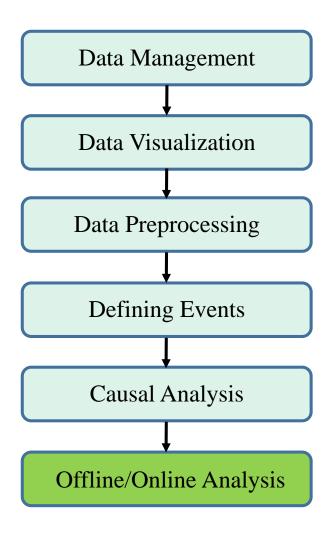




#### Online monitoring (this simulates the implementation of online process monitoring)

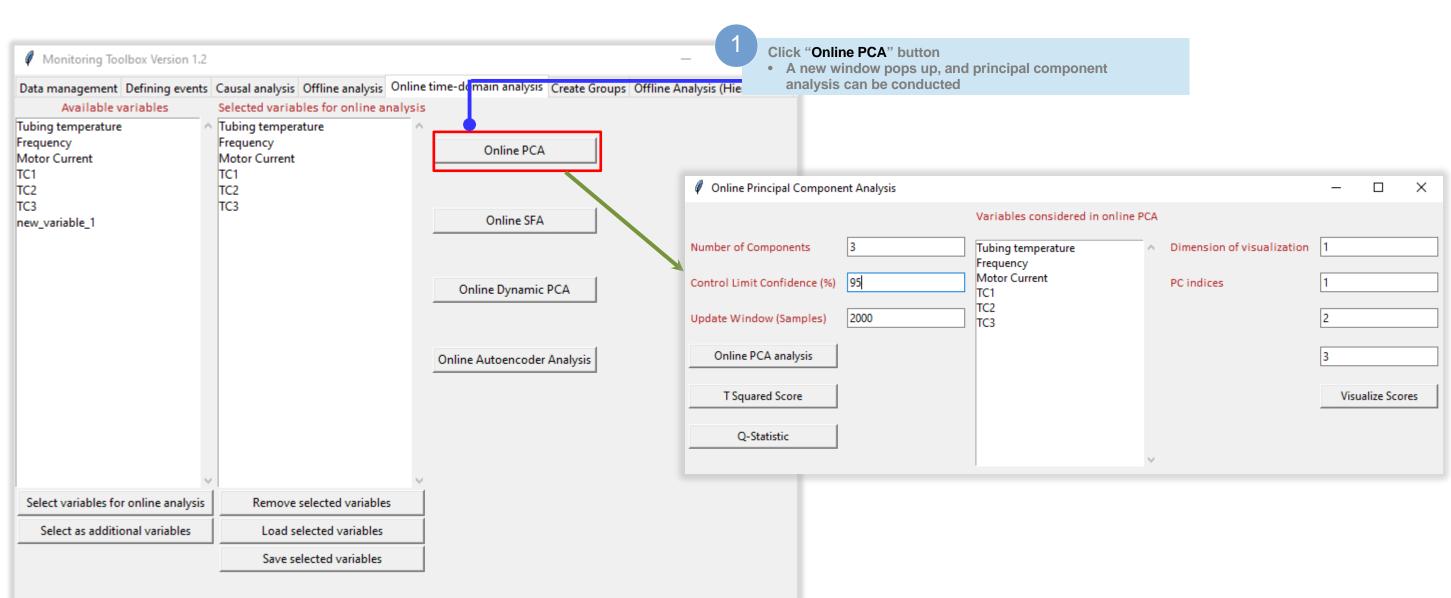
Under the "Online analysis" tab, four algorithms are provided and can be used to monitor the process and detect events





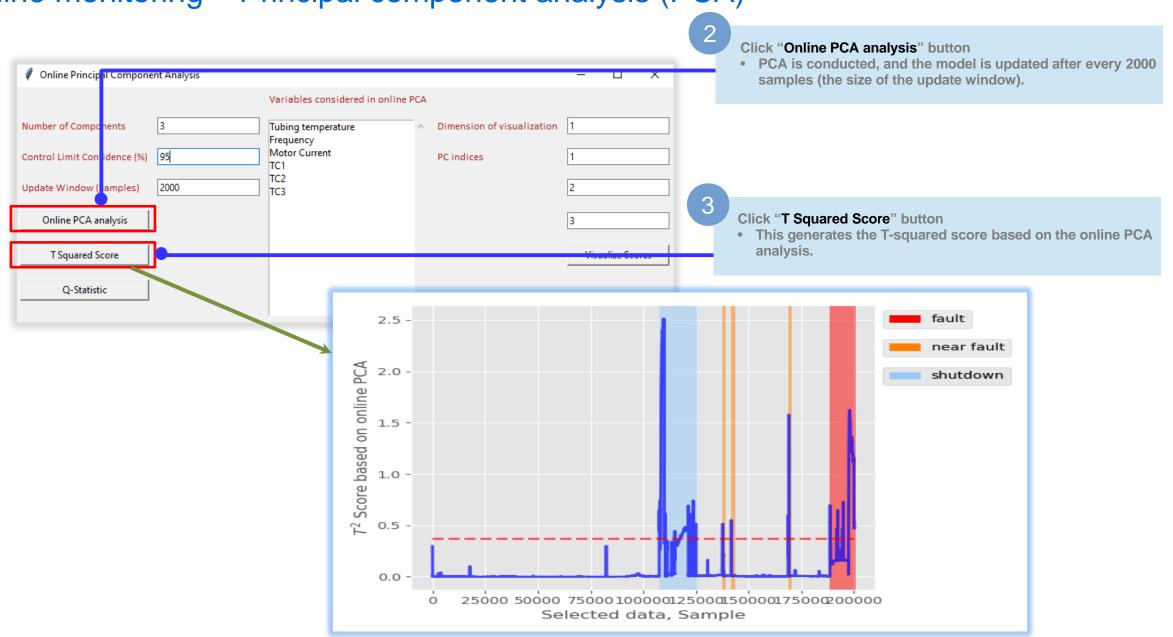


#### Online monitoring – Principal component analysis





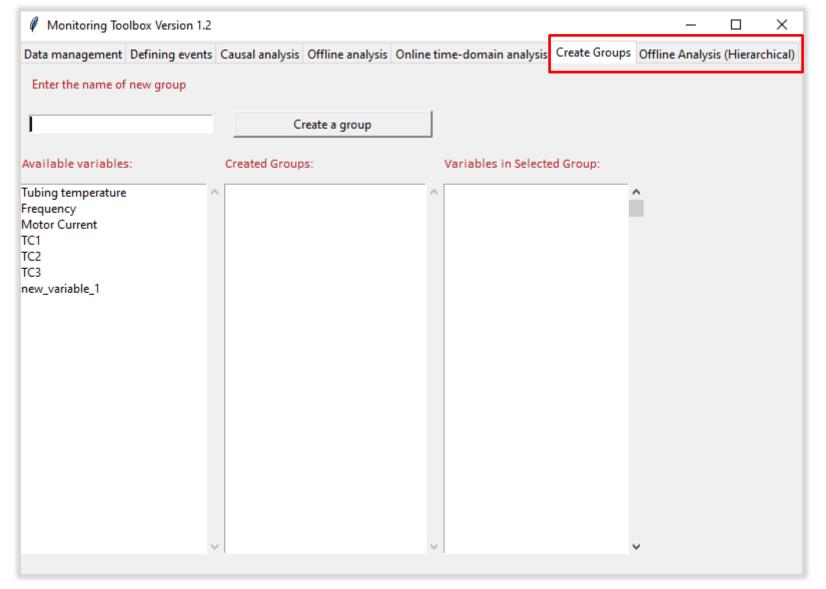
Online monitoring – Principal component analysis (PCA)

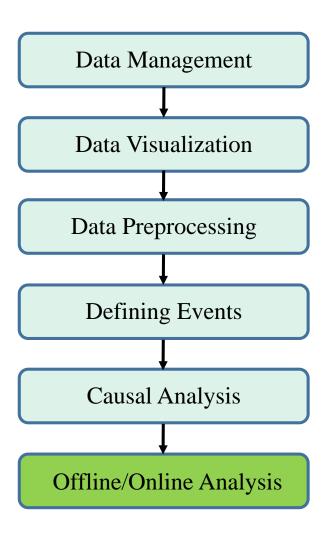




#### Hierarchical distributed monitoring (this is favorable when the number of variables is large)

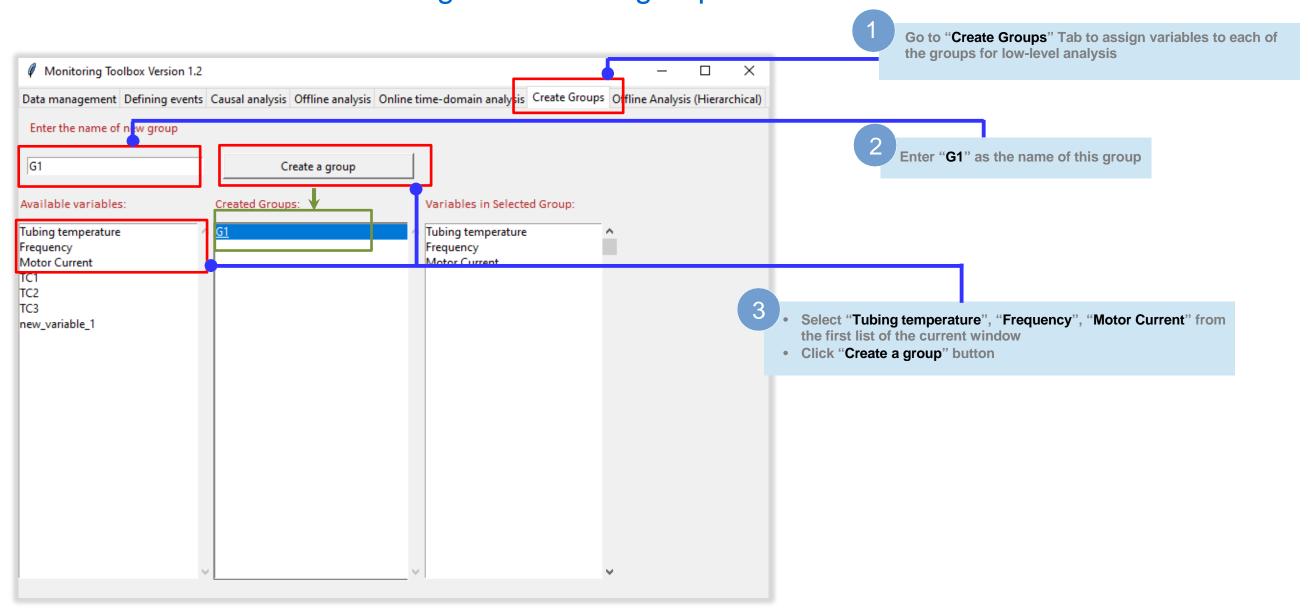
Both "Create Groups" and "Offline Analysis (Hierarchical)" tabs are used.





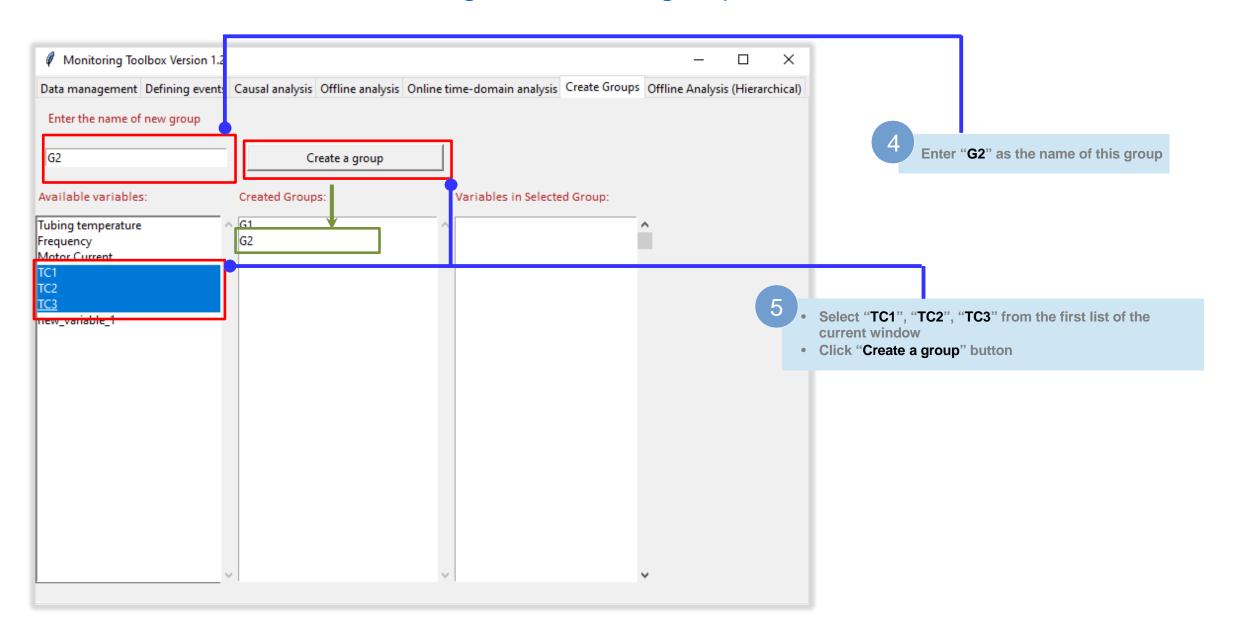


Hierarchical distributed monitoring – Construct groups of variables



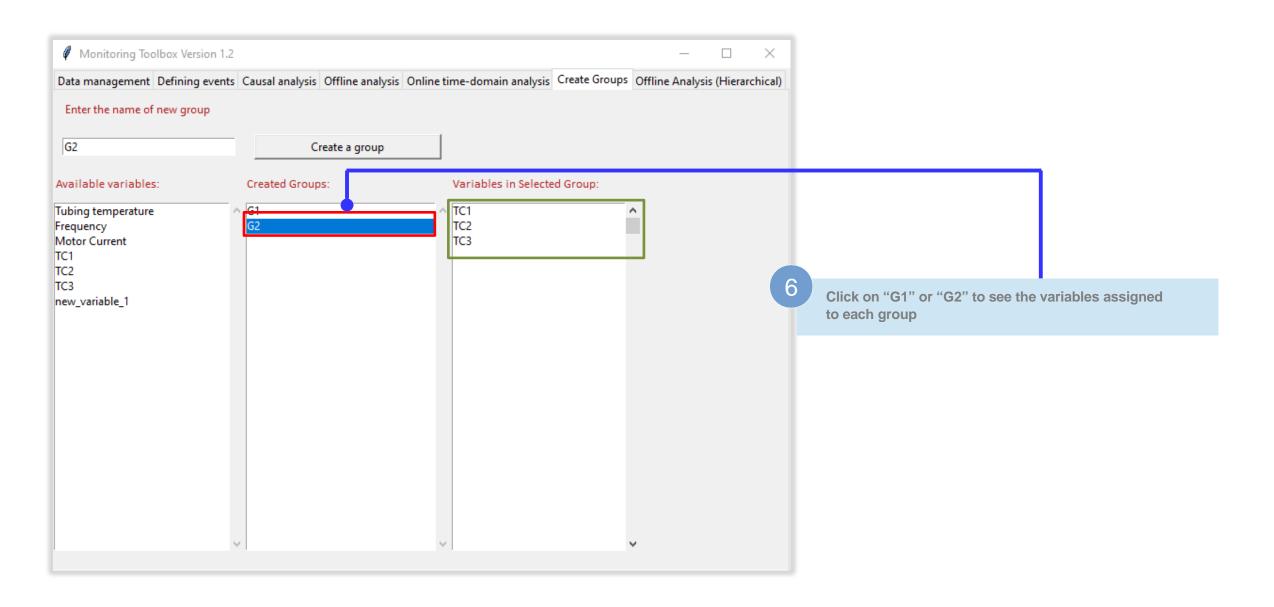


#### Hierarchical distributed monitoring – Construct groups of variables



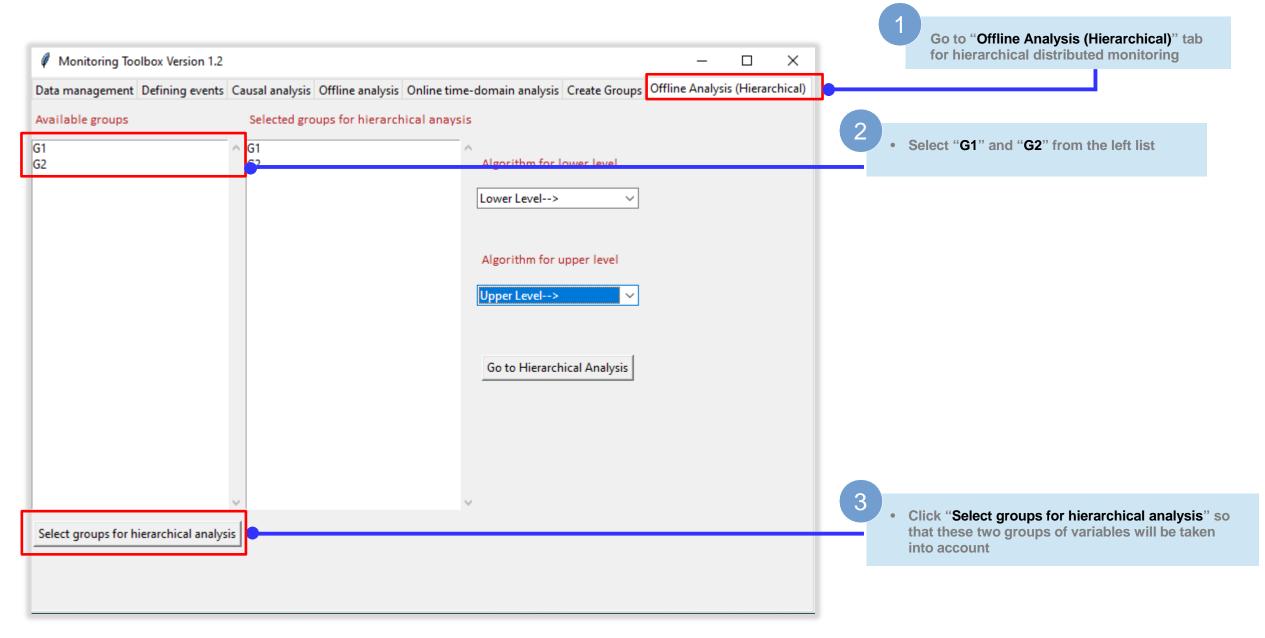


Hierarchical distributed monitoring – Construct groups of variables



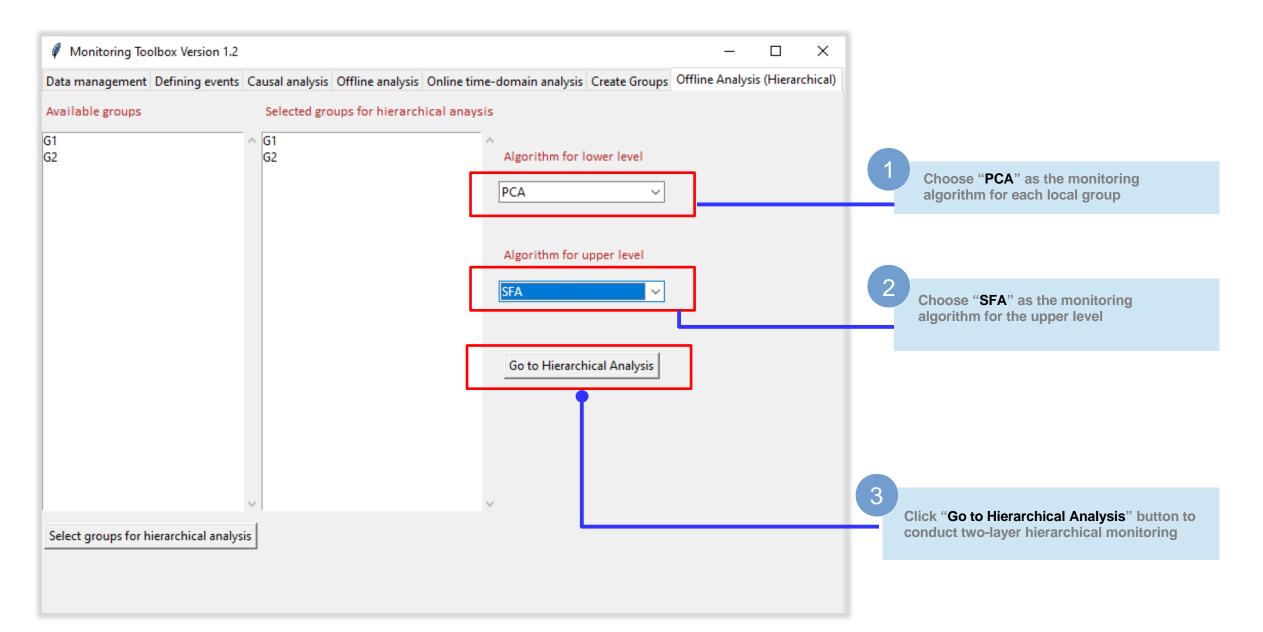


Hierarchical distributed monitoring – Monitoring analysis





#### Hierarchical distributed monitoring – Monitoring analysis





#### Hierarchical distributed monitoring – Monitoring analysis

